

J -Hermitian determinantal point processes: balanced rigidity and balanced Palm equivalence

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Abstract

We study Palm measures of determinantal point processes with J -Hermitian correlation kernels. A point process \mathbb{P} on the punctured real line $\mathbb{R}^* = \mathbb{R}_+ \sqcup \mathbb{R}_-$ is said to be *balanced rigid* if for any precompact subset $B \subset \mathbb{R}^*$, the *difference* between the numbers of particles of a configuration inside $B \cap \mathbb{R}_+$ and $B \cap \mathbb{R}_-$ is almost surely determined by the configuration outside B . The point process \mathbb{P} is said to have the *balanced Palm equivalence property* if any reduced Palm measure conditioned at $2n$ distinct points, n in \mathbb{R}_+ and n in \mathbb{R}_- , is equivalent to the \mathbb{P} .

We formulate general criteria for determinantal point processes with J -Hermitian correlation kernels to be balanced rigid and to have the balanced Palm equivalence property and prove, in particular, that the determinantal point processes with Whittaker kernels of Borodin and Olshanski are balanced rigid and have the balanced Palm equivalence property.

Keywords. Determinantal point processes; J -Hermitian kernel; Whittaker kernels; L -processes; Palm measures; balanced rigidity; balanced Palm equivalence property.

Contents

1	Introduction	2
1.1	Palm measures of determinantal point processes	2
1.2	Main results for Whittaker kernels	3
1.3	Main results for general J -Hermitian kernels	6

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1.3.1	Theorem A for general J -Hermitian kernels	7
1.3.2	Theorem B for general J -Hermitian kernels	9
1.4	Olshanski's Problem	11
1.5	Organization of the paper and schemes of proofs	14
2	Preliminaries	15
2.1	Palm measures of determinantal point processes	16
2.2	J -Hermitian kernels and extended Fredholm determinants	18
3	Balanced rigidity	21
3.1	A sufficient condition for balanced rigidity	21
3.2	Proof of Theorem A	26
4	Balanced Palm equivalence property	28
4.1	Palm measures of L -processes	29
4.2	Sufficient condition for equivalence of two L -processes	36
4.3	Remarks on Conditions 3, 4, 5 and Proofs of Lemmas 4.6, 4.7, 4.8	37
4.4	A preliminary version of Theorem 4.9	40
4.5	Regularization of additive and multiplicative functionals	42
4.5.1	Additive functionals	42
4.5.2	Multiplicative functionals	45
4.6	Proof of Theorem 4.14	46
4.7	Proof of Proposition 4.15	47
4.8	Proof of Theorem 4.9	52
4.9	Proof of Theorem B	52
5	Appendix	54

1 Introduction

1.1 Palm measures of determinantal point processes

The present paper is the first one devoted to the equivalence and mutual singularity relations between reduced Palm measures of determinantal point processes with J -Hermitian correlation kernels.

As a concrete model, we consider the family of determinantal point processes on the punctured real line $\mathbb{R}^* = \mathbb{R} \setminus \{0\}$ with Whittaker kernels of Borodin and Olshanski [Bor, Ols], scaling limits of the so-called z -measures of partitions [BO00, BO05]. For these determinantal point processes, we observe a new effect: the reduced Palm measure conditioned at $2n$ points, n on the positive, n on the negative semi-axis, is equivalent to

the initial determinantal measure; while if $n \neq k$, then the initial measure and the reduced Palm measure conditioned at $n + k$ points, n on the positive, k on the negative semi-axis, are mutually singular. In the former case, the Radon-Nikodym derivatives between the reduced Palm measures and the initial determinantal measure are found explicitly as *regularized multiplicative functionals*.

In the case of determinantal measures with kernels given by Hermitian projection operators, the statement that two such measures differ by a multiplicative functional can be checked on the level of the corresponding subspaces, the ranges of our projections: in fact, it suffices to verify that these subspaces differ by multiplication by a function, see [Buf12, Buf13] for precise statements.

Although J -Hermitian operators considered in this paper are closely related to certain Hermitian projection operators, it does not seem possible to work with their ranges. Instead, we use the fact that the determinantal point processes with the Whittaker kernels admit so-called L -kernels. Following Borodin and Olshanski, such processes will be called L -processes. Two L -processes differ by a multiplicative functional once corresponding L -kernels themselves differ by multiplication by a function on the left and on the right.

The realization of this scheme requires some effort. First, Palm measures of an L -process, generally speaking, do not admit an L -kernel (this can be seen already on the level of discrete phase spaces: indeed, Borodin and Rains [BR05] shown that any determinantal point process can be obtained from an L -process by conditioning). Second, in developing the formalism of the regularized multiplicative functionals, we are not able to use the standard linear statistics as in [Buf14, BQ]. We use the *twisted* ones instead (see (3.28) and (4.84) below for the definitions); in particular, an extended version of Fredholm determinants is used.

1.2 Main results for Whittaker kernels

We start by formulating our main results for a concrete model: the family of the determinantal point processes with Whittaker kernels of Borodin and Olshanski. The reader is referred to [BO00, Bor, Ols] for the origin of these point processes in the problem of harmonic analysis on the infinite symmetric group and to §2 below for a reminder of the main definitions related to determinantal point processes.

Let $\mathbb{R}^* = \mathbb{R} \setminus \{0\}$ be the punctured real line. By a *configuration* on \mathbb{R}^* , we mean a *locally finite* subset $X \subset \mathbb{R}^*$, that is, X is a subset of \mathbb{R}^* such that for any compact subset $B \subset \mathbb{R}^*$, the cardinality $\#(X \cap B)$ of the intersection of the subsets X and B is finite. Define the *space of configurations* on \mathbb{R}^* by

$$\text{Conf}(\mathbb{R}^*) := \{X \subset \mathbb{R}^* : \text{for any compact subset } B \subset \mathbb{R}^*, \#(X \cap B) < \infty\}.$$

The space of configurations $\text{Conf}(\mathbb{R}^*)$ is naturally equipped with a Borel structure, see §2. A point process on \mathbb{R}^* is by definition a Borel probability on $\text{Conf}(\mathbb{R}^*)$.

The family of the determinantal point processes with Whittaker kernels of Borodin and Olshanski is a 2-parameter family $\mathcal{P}_{z,z'}$ ⁴ of determinantal point processes on \mathbb{R}^* . The two parameters $z, z' \in \mathbb{C}$ satisfy one of the following conditions:

- either $z' = \bar{z}$ and $z \in \mathbb{C} \setminus \mathbb{Z}$,
- or $z, z' \in \mathbb{R}$ and there exists $m \in \mathbb{Z}$ such that $m < z, z' < m + 1$.

Following [BO00, formula (5.6)], we now write the correlation kernel of the determinantal point process $\mathcal{P}_{z,z'}$ explicitly. Fix two parameters $z, z' \in \mathbb{C}$ such that one of the two conditions as above is satisfied. Set

$$\begin{aligned} \mathcal{P}_{\pm}(x) &= \frac{(zz')^{1/4}}{(\Gamma(1 \pm z)\Gamma(1 \pm z')x)^{1/2}} W_{\frac{\pm(z+z')+1}{2}, \frac{z-z'}{2}}(x), \\ \mathcal{Q}_{\pm}(x) &= \frac{(zz')^{3/4}}{(\Gamma(1 \pm z)\Gamma(1 \pm z')x)^{1/2}} W_{\frac{\pm(z+z')-1}{2}, \frac{z-z'}{2}}(x), \end{aligned} \quad (1.1)$$

where $\Gamma(\cdot)$ is the Euler Gamma-function and $W_{a,b}(\cdot)$ is the Whittaker function with parameter $a, b \in \mathbb{C}$, see [EMOT53, 6.9] for the definition of Whittaker functions. The correlation kernel of $\mathcal{P}_{z,z'}$ is given by

$$\mathcal{K}_{z,z'}(x, y) = \begin{cases} \frac{\mathcal{P}_+(x)\mathcal{Q}_+(y) - \mathcal{Q}_+(x)\mathcal{P}_+(y)}{x - y}, & \text{for } x > 0, y > 0; \\ \frac{\mathcal{P}_+(x)\mathcal{P}_-(-y) + \mathcal{Q}_+(x)\mathcal{Q}_-(-y)}{x - y}, & \text{for } x > 0, y < 0; \\ \frac{\mathcal{P}_-(-x)\mathcal{P}_+(y) + \mathcal{Q}_-(-x)\mathcal{Q}_+(y)}{x - y}, & \text{for } x < 0, y > 0; \\ \frac{\mathcal{P}_-(-x)\mathcal{Q}_-(-y) - \mathcal{Q}_-(-x)\mathcal{P}_-(-y)}{y - x}, & \text{for } x < 0, y < 0. \end{cases} \quad (1.2)$$

These kernels $\mathcal{K}_{z,z'}$ are called Whittaker kernels.

Recall that given a finite set S , we denote its cardinality by $\#(S)$. Denote by \mathbb{R}_+ the positive semi-axis and \mathbb{R}_- the negative semi-axis. Our first main result, in case of Whittaker kernel model, is

Theorem A. Assume that the parameters (z, z') are such that $z' = \bar{z}$ and $z \in \mathbb{C} \setminus \mathbb{R}$. Then for any subset $B \subset \mathbb{R}^*$ having a positive distance from the origin, the difference

$$\#(B \cap X \cap \mathbb{R}_+) - \#(B \cap X \cap \mathbb{R}_-)$$

⁴It was denoted as $\tilde{\mathcal{P}}_{z,z'}$ in [BO00].

is $\mathcal{P}_{z,z'}$ -almost surely determined by $X \cap (\mathbb{R}^* \setminus B)$, the configuration outside B . That is, there exists a measurable function $N_B^{out} : \text{Conf}(\mathbb{R}^*) \rightarrow \mathbb{Z}$, such that for $\mathcal{P}_{z,z'}$ -almost every configuration $X \in \text{Conf}(\mathbb{R}^*)$, we have

$$\#(B \cap X \cap \mathbb{R}_+) - \#(B \cap X \cap \mathbb{R}_-) = N_B^{out}(X \setminus B).$$

In particular, if $B \subset \mathbb{R}_+$ is a subset in the positive semi-axis with a positive distance from the origin, then $\#(B \cap X)$ is $\mathcal{P}_{z,z'}$ -almost surely determined by $X \cap (\mathbb{R}^* \setminus B)$. If B is in the negative semi-axis, the same result holds.

Remark 1.1. When the subset B is either in positive semi-axis or in negative semi-axis, we recover the usual *number rigidity property* of Ghosh [Gho14], Ghosh and Peres [GP].

If \mathbb{P} is a point process on \mathbb{R}^* and if $\mathbf{p} = (p_1, \dots, p_m) \in (\mathbb{R}^*)^m$ is an m -tuple of distinct points in \mathbb{R}^* , then we denote $\mathbb{P}^{\mathbf{p}}$ the *reduced Palm measure* of \mathbb{P} conditioned at the points p_1, \dots, p_m . See §2.1 for the formal definition of the reduced Palm measures.

Using a variant of Proposition 8.1 in [BQ], we derive from Theorem A the following corollary.

Corollary 1.1. *Assume that the parameters (z, z') are such that $z' = \bar{z}$ and $z \in \mathbb{C} \setminus \mathbb{R}$. Let n, k be two non-negative integers such that $n \neq k$. Then for Lebesgue-almost every $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_k^-) \in \mathbb{R}_+^n \times \mathbb{R}_-^k$ of distinct points, the reduced Palm measure $\mathcal{P}_{z,z'}^{\mathbf{p}}$ and the initial determinantal measure $\mathcal{P}_{z,z'}$ are mutually singular.*

We now proceed to formulating our second main result which gives equivalence of the reduced Palm measures $\mathcal{P}_{z,z'}^{\mathbf{p}}$ and $\mathcal{P}_{z,z'}$, under the conditions that the parameters $z, z' \in \mathbb{C}$ are such that $|z + z'| < 1$ and $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-) \in \mathbb{R}_+^n \times \mathbb{R}_-^n$ is a $2n$ -tuple of distinct points in \mathbb{R}^* with equal numbers of points from positive and negative semi-axis. The Radon-Nikodym derivative $d\mathcal{P}_{z,z'}^{\mathbf{p}}/d\mathcal{P}_{z,z'}$ is computed explicitly.

We start with an auxiliary proposition

Proposition 1.2. *Assume that the two parameters $z, z' \in \mathbb{C}$ are such that $|z + z'| < 1$. Then the following limit*

$$\begin{aligned} \bar{S}_{\mathbf{p}}(X) := & \lim_{\delta \rightarrow 0^+} \left\{ \sum_{i=1}^n \left(\sum_{x \in X \cap (\delta, \infty)} \log \left| \frac{x/p_i^+ - 1}{x/p_i^- - 1} \right|^2 - \sum_{x \in X \cap (-\infty, -\delta)} \log \left| \frac{x/p_i^- - 1}{x/p_i^+ - 1} \right|^2 \right) \right. \\ & \left. - \mathbb{E}_{\mathcal{P}_{z,z'}} \sum_{i=1}^n \left(\sum_{x \in X \cap (\delta, \infty)} \log \left| \frac{x/p_i^+ - 1}{x/p_i^- - 1} \right|^2 - \sum_{x \in X \cap (-\infty, -\delta)} \log \left| \frac{x/p_i^- - 1}{x/p_i^+ - 1} \right|^2 \right) \right\} \end{aligned}$$

exists for $\mathcal{P}_{z,z'}$ -almost every configuration $X \in \text{Conf}(\mathbb{R}^*)$. Moreover, we have

$$\exp(\bar{S}_{\mathbf{p}}) \in L^1(\text{Conf}(\mathbb{R}^*), \mathcal{P}_{z,z'}).$$

Theorem B. Assume that the two parameters $z, z' \in \mathbb{C}$ are such that $|z + z'| < 1$. Then the determinantal point process $\mathcal{P}_{z,z'}$ possesses the following property: for Lebesgue almost every $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-) \in \mathbb{R}_+^n \times \mathbb{R}_-^n$ of distinct points, the reduced Palm measure $\mathcal{P}_{z,z'}^{\mathbf{p}}$ is equivalent to the initial determinantal measure $\mathcal{P}_{z,z'}$. For the Radon-Nikodym derivative, we have the $\mathcal{P}_{z,z'}$ -almost sure equality

$$\frac{d\mathcal{P}_{z,z'}^{\mathbf{p}}}{d\mathcal{P}_{z,z'}}(X) = \frac{\exp(\bar{S}_{\mathbf{p}}(X))}{\mathbb{E}_{\mathcal{P}_{z,z'}}[\exp(\bar{S}_{\mathbf{p}})]}.$$

Remark 1.2. In [Ols11], the determinantal point processes $\mathcal{P}_{z,z'}$ are obtained as scaling limits of determinantal point processes on $\text{Conf}(\mathbb{Z}')$ with hypergeometric kernel measures, where $\mathbb{Z}' = \frac{1}{2} + \mathbb{Z}$ is the set of half-integers. Being point processes all concentrate on the set of *balanced configurations* with a finite number of particles:

$$\{X \in \text{Conf}(\mathbb{Z}') : \#(X \cap \mathbb{Z}'_+) = \#(X \cap \mathbb{Z}'_-) < \infty\},$$

the determinantal point processes with hypergeometric kernels are of course *balanced rigid* in the sense of Theorem A. However, as already shown in Hermitian kernel case, the rigidity property is not stable under taking limits. Indeed, orthogonal polynomial ensembles, having a fixed number of particles, are rigid in the sense of Ghosh [Gho14] and Ghosh-Peres [GP], while in general this is not the case for their scaling limits. For example, as Holroyd and Soo [HS13] showed, the determinantal point process on the unit disk \mathbb{D} with Bergman kernel:

$$K_{\text{Berg}}(z, w) = \frac{1}{\pi(1 - z\bar{w})^2}, \quad z, w \in \mathbb{D},$$

is not rigid (the Radon-Nikodym derivatives between this measure and its Palm measures are computed in [BQ]), but is nonetheless the limit of the following sequence of rigid determinantal point processes whose kernels are given by finite rank orthogonal projections:

$$K_{\text{Berg}}^{(n)}(z, w) = \frac{1}{\pi} \sum_{k=0}^{n-1} (k+1)(z\bar{w})^k, \quad z, w \in \mathbb{D}.$$

1.3 Main results for general J -Hermitian kernels

Our proofs of Theorem A and Theorem B do not proceed by limit transition from processes with finitely many particles and work for more general J -Hermitian kernels.

We briefly recall the necessary definitions. Let P_+, P_- denote the orthogonal projections on $L^2(\mathbb{R}) = L^2(\mathbb{R}, dx)$ whose ranges are the subspaces $L^2(\mathbb{R}_+)$ and $L^2(\mathbb{R}_-)$ respectively. Define a bounded linear operator J on $L^2(\mathbb{R})$ by

$$J := P_+ - P_-.$$

Introduce an indefinite J -scalar product $[\cdot, \cdot]$ on $L^2(\mathbb{R})$ by the formula

$$[f, g] := (Jf, g) = (P_+f, P_+g) - (P_-f, P_-g), \quad f, g \in L^2(\mathbb{R}),$$

where (\cdot, \cdot) denotes the usual scalar product in $L^2(\mathbb{R})$. A bounded linear operator K on $L^2(\mathbb{R})$ is called J -self-adjoint if $[Kf, g] = [f, Kg]$ for any pair $f, g \in L^2(\mathbb{R})$. By slightly abusing the notation, the kernel of an integral operator K will denote again by K . A kernel K is called a J -Hermitian kernel, if the corresponding operator K is J -self-adjoint. More precisely, $K(x, y)$ is J -Hermitian if it induces a bounded linear operator and if

$$K(x, y) = \operatorname{sgn}(x)\operatorname{sgn}(y)\overline{K(y, x)}, \quad x, y \in \mathbb{R}^*, \quad (1.3)$$

where $\operatorname{sgn}(x)$ is the sign of the real number $x \in \mathbb{R}^*$.

By convention, a bounded measurable function $f : \mathbb{R}^* \rightarrow \mathbb{C}$ will be identified with the bounded linear operator M_f on $L^2(\mathbb{R})$ defined by

$$M_f(g) = fg, \text{ for any } g \in L^2(\mathbb{R}).$$

The notation fK (or $f \cdot K$) and Kf (or $K \cdot f$) stands for the composition operators $M_f \circ K$ and $K \circ M_f$ respectively.

Given a bounded linear operator K on $L^2(\mathbb{R})$, we set

$$\widehat{K} := \operatorname{sgn} \cdot K + \chi_{\mathbb{R}_-} = P_+K + P_-(1 - K). \quad (1.4)$$

An operator K is J -self-adjoint if and only if the operator \widehat{K} is self-adjoint in the usual sense.

The following Theorem of Lytvynov gives a necessary and sufficient condition for the existence of a determinantal point process with a given J -Hermitian kernel.

Theorem 1.3 (E. Lytvynov [Lyt13]). *Let K be a J -self-adjoint bounded linear operator on $L^2(\mathbb{R})$. Assume that the operators P_+KP_+ and P_-KP_- are non-negative. Assume also that, for any bounded subsets Δ_1, Δ_2 of \mathbb{R} such that $\Delta_1 \subset \mathbb{R}_+$ and $\Delta_2 \subset \mathbb{R}_-$, the operators $\chi_{\Delta_i}K\chi_{\Delta_i}$ ($i = 1, 2$) are in trace-class, while $\chi_{\Delta_2}K\chi_{\Delta_1}$ is Hilbert-Schmidt. Then the integral kernel $K(x, y)$ of the operator K is the correlation kernel of a determinantal point process on \mathbb{R} if and only if $0 \leq \widehat{K} \leq 1$.*

The determinantal point process induced by a correlation kernel K as in Theorem 1.3 will be denoted by \mathbb{P}_K .

1.3.1 Theorem A for general J -Hermitian kernels

We now formulate a general variant of Theorem A, namely, a sufficient condition on the J -Hermitian kernel K for the determinantal point process \mathbb{P}_K to be balanced rigid.

For the purpose of our later use of Fourier analysis, we state our result in this part only for determinantal point processes with a singularity at infinity. The case with a single singularity at the origin, such as the determinantal point processes with Whittaker kernels, can be easily transformed to this case by the change of variables $x \mapsto 1/x$.

We need the following two conditions on the kernel K : Condition 1 guarantees that the J -Hermitian kernel K is indeed a correlation kernel of a certain determinantal point process and the variance of the linear statistics

$$\sum_{x \in X} \text{sgn}(x) \varphi(x), \quad (\text{where } X \in \text{Conf}(\mathbb{R})),$$

can be expressed by a simple formula, see Lemma 3.2 below. Condition 2 guarantees that the diagonal coefficient $K(x, x)$ is locally integrable on \mathbb{R} and controls the rate of decay of off-diagonal coefficients $K(x, y)$ when $|x - y|$ is large. The former condition on $K(x, x)$ implies in particular that the associated determinantal point process has no accumulation point at any point of the real line.

Condition 1. Assume that K is the integral kernel of a bounded linear operator on $L^2(\mathbb{R})$ such that

- the operators $P_+ K P_+$ and $P_- K P_-$ are non-negative. Moreover, for any bounded subsets Δ_1, Δ_2 of the real line such that $\Delta_1 \subset \mathbb{R}_+$ and $\Delta_2 \subset \mathbb{R}_-$, the operators $\chi_{\Delta_i} K \chi_{\Delta_i}$ ($i = 1, 2$) are in trace-class, and the operator $\chi_{\Delta_2} K \chi_{\Delta_1}$ is Hilbert-Schmidt.
- the following operator

$$\widehat{K} := \text{sgn} \cdot K + \chi_{\mathbb{R}_-}$$

defines an *orthogonal projection* on $L^2(\mathbb{R}, dx)$.

Condition 2. Fix $M > 0$. Assume that the kernel K satisfies the following conditions:

- Let $K(x, x)$ be the diagonal value of the kernel K , then for any $R > 0$, we have

$$\int_{|x| \leq R} K(x, x) dx < \infty; \tag{1.5}$$

- There exists a non-negative integrable function $\Phi \in L^1(\mathbb{R}, dt)$ satisfying

$$\int_{|t| \geq R} \Phi(t) dt = O(R^{-1}) \text{ as } R \rightarrow \infty, \tag{1.6}$$

such that if $|x| \geq M, |y| \geq M$, then

$$|K(x, y)|^2 \leq \Phi(x - y); \tag{1.7}$$

The general variant of Theorem A is

Theorem 1.4 (A sufficient condition for balanced rigidity). *Let K be a J -Hermitian kernel satisfying Conditions 1 and 2. Then the determinantal point process \mathbb{P}_K possesses the following rigid property: for any bounded Borel subset $B \subset \mathbb{R}$, there exists a measurable function $N_B^{out} : \text{Conf}(\mathbb{R}) \rightarrow \mathbb{Z}$, such that for \mathbb{P}_K -almost every $X \in \text{Conf}(\mathbb{R})$, we have*

$$\#(B \cap X \cap \mathbb{R}_+) - \#(B \cap X \cap \mathbb{R}_-) = N_B^{out}(X \setminus B).$$

In particular, if $B \subset \mathbb{R}_+$ is a bounded subset in the positive semi-axis, then $\#(B \cap X)$ is \mathbb{P}_K -almost surely determined by $X \cap (\mathbb{R} \setminus B)$. If $B \subset \mathbb{R}_-$ is a bounded subset in the negative semi-axis, the same result holds.

Using a variant of Proposition 8.1 in [BQ], we derive from Theorem 1.4 the following corollary.

Corollary 1.5. *Let K be a J -Hermitian kernel satisfying Conditions 1 and 2. Let n, k be two non-negative integers such that $n \neq k$. Then for Lebesgue-almost every $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_k^-) \in \mathbb{R}_+^n \times \mathbb{R}_-^k$ of distinct points, the reduced Palm measure $\mathbb{P}_K^{\mathbf{p}}$ and the initial determinantal measure \mathbb{P}_K are mutually singular.*

1.3.2 Theorem B for general J -Hermitian kernels

We now formulate a general variant of Theorem B, namely, a sufficient condition for the determinantal point process to have *balanced Palm equivalence property* in the sense of Theorem B, see also Definition 4.1 below. In this part, let us state the result in the case where there is a single singularity at the origin (rather than a singularity at infinity).

We first need the definition of L -processes of Borodin and Olshanski.

Definition 1.1 (L -kernel). Given a bounded linear operator K on $L^2(\mathbb{R})$, if $1 - K$ is invertible, then we define the L -operator of K by

$$L = K(1 - K)^{-1}.$$

In order to emphasize that the operator K depends on L , we will sometimes write K_L instead of K , thus having

$$K_L = L(1 + L)^{-1}.$$

Condition 3. Assume that L is a bounded operator on $L^2(\mathbb{R}^*, dx)$ having the following block form:

$$L(x, y) = \begin{bmatrix} 0 & V \\ -V^* & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{A^+(x)A^-(y)}{x-y} \\ \frac{A^-(x)A^+(y)}{x-y} & 0 \end{bmatrix}, \quad (1.8)$$

where $A \in C^1(\mathbb{R}^*) \cap L^2(\mathbb{R}^*, dx)$ is a real-valued function and

$$A^+ := A\chi_{\mathbb{R}_+} \text{ and } A^- := A\chi_{\mathbb{R}_-}.$$

In other words, the operator L admits an integral kernel given by

$$L(x, y) = \frac{A^+(x)A^-(y) + A^-(x)A^+(y)}{x - y}, \quad x, y \in \mathbb{R}^*. \quad (1.9)$$

We will also assume that the support of the function A is the whole punctured line \mathbb{R}^* .

Lemma 1.6. *Let L be an operator satisfying Condition 3. Then the operator K_L is a J -self-adjoint operator satisfying all the conditions of the Lytvynov's Theorem 1.3. In particular, K_L is the correlation kernel of a determinantal point process on \mathbb{R}^* .*

Definition 1.2 (L -processes). We denote by μ_L the determinantal point process on \mathbb{R}^* whose correlation kernel is $K_L = L(1 + L)^{-1}$, that is

$$\mu_L := \mathbb{P}_{K_L}. \quad (1.10)$$

Following Borodin and Olshanski, such processes will be called L -processes.

We need the following auxiliary propositions.

Let $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-)$ be a $2n$ -tuple of real numbers such that $p_i^+ > 0$ and $p_i^- < 0$ for $i = 1, \dots, n$. Moreover, assume that $p_i^+ \neq p_j^+, p_i^- \neq p_j^-$ when $i \neq j$. Define

$$f_{\mathbf{p}}(x) = \prod_{i=1}^n \left(\frac{x/p_i^+ - 1}{x/p_i^- - 1} \chi_{\mathbb{R}_+}(x) + \frac{x/p_i^- - 1}{x/p_i^+ - 1} \chi_{\mathbb{R}_-}(x) \right). \quad (1.11)$$

Proposition 1.7. *Let L be an operator satisfying Condition 3. Let μ_L be the determinantal point process on \mathbb{R}^* whose correlation kernel is $K_L = L(1 + L)^{-1}$. Then the reduced Palm measure $\mu_L^{\mathbf{p}}$ conditioned at a $2n$ -tuple of distinct points*

$$\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-) \in \mathbb{R}_+^n \times \mathbb{R}_-^n, \quad (1.12)$$

is again an L -process and is given by

$$\mu_L^{\mathbf{p}} = \mu_{f_{\mathbf{p}} L f_{\mathbf{p}}}. \quad (1.13)$$

Proposition 1.8. *Let L be an operator satisfying Condition 3 and let \mathbf{p} be a $2n$ -tuple of distinct points in \mathbb{R}^* given as in (1.12), the function $f_{\mathbf{p}}$ is defined by formula (1.11). Then the following limit*

$$\begin{aligned} \overline{S}_{\mathbf{p}}(X) := & \lim_{\delta \rightarrow 0^+} \left\{ \sum_{i=1}^n \left(\sum_{x \in X \cap (\delta, \infty)} \log \left| \frac{x/p_i^+ - 1}{x/p_i^- - 1} \right|^2 - \sum_{x \in X \cap (-\infty, -\delta)} \log \left| \frac{x/p_i^- - 1}{x/p_i^+ - 1} \right|^2 \right) \right. \\ & \left. - \mathbb{E}_{\mu_L} \sum_{i=1}^n \left(\sum_{x \in X \cap (\delta, \infty)} \log \left| \frac{x/p_i^+ - 1}{x/p_i^- - 1} \right|^2 - \sum_{x \in X \cap (-\infty, -\delta)} \log \left| \frac{x/p_i^- - 1}{x/p_i^+ - 1} \right|^2 \right) \right\} \end{aligned}$$

exists for μ_L -almost every configuration $X \in \text{Conf}(\mathbb{R}^*)$. Moreover, we have

$$\exp(\overline{S}_{\mathfrak{p}}) \in L^1(\text{Conf}(\mathbb{R}^*), \mu_L).$$

Theorem 1.9 (A sufficient condition to have balanced Palm equivalence property). *Let L be an operator satisfying Condition 3 and let \mathfrak{p} be a $2n$ -tuple of distinct points in \mathbb{R}^* given as in (1.12), let $f_{\mathfrak{p}}$ denote the function defined by formula (1.11). Then the reduced Palm measure $\mu_L^{\mathfrak{p}}$ is equivalent to μ_L . For the Radon-Nikodym derivative, we have the μ_L -almost sure equality:*

$$\frac{d\mu_L^{\mathfrak{p}}}{d\mu_L}(X) = \frac{\exp(\overline{S}_{\mathfrak{p}}(X))}{\mathbb{E}_{\mu_L}[\exp(\overline{S}_{\mathfrak{p}})]}.$$

1.4 Olshanski's Problem

Olshanski [Ols11] posed the following

Problem. *Let P_1 and P_2 be two determinantal point processes on a common phase space with correlation kernels $K_1(x, y)$ and $K_2(x, y)$ respectively. Decide the equivalence and the mutual singularity relations between P_1 and P_2 by inspection of their correlation kernels $K_1(x, y)$ and $K_2(x, y)$. When P_1 and P_2 are equivalent, calculate the Radon-Nikodym derivative between them.*

We now briefly mention the previous works on this problem for projection kernels and note the particle-hole duality relation, in the case of discrete phase spaces, of these results to the results of the present paper.

• The Gamma-kernel.

Olshanski [Ols11] obtained the quasi-invariance of the so-called Gamma kernel determinantal point processes on the space \mathbb{Z}' of half-integers under the action of the group $S(\infty)$ of finite permutations of arbitrary size. The group $S(\infty)$ acts naturally on \mathbb{Z}' and hence on the space of configurations over \mathbb{Z}' . Let P_1 be the Gamma kernel determinantal point process on \mathbb{Z}' with the correlation kernel denoted by K_1 , see [BO05] for the precise definition. Take an element $\sigma \in S(\infty)$, denote $P_2 = \sigma_*(P_1)$ the determinantal point process obtained by the transformation σ on the space of configurations $\text{Conf}(\mathbb{Z}') = 2^{\mathbb{Z}'}$. Then P_2 has a correlation kernel given by

$$K_2(x, y) = K_1(\sigma^{-1}(x), \sigma^{-1}(y)) \quad x, y \in \mathbb{Z}'.$$

By *limit transition* from finite particle systems, Olshanski proved the equivalence of P_1 and P_2 and calculated the Radon-Nikodym derivative between them as a multiplicative functional.

- **Determinantal point processes with integrable kernels.**

In [Buf14], it has been proved that all determinantal point processes on \mathbb{R} (or \mathbb{Z}) whose correlation kernels K are Hermitian orthogonal projections and have an integrable form as follows:

$$K(x, y) = \frac{A(x)B(y) - A(y)B(x)}{x - y}, \quad x, y \in \mathbb{R} \text{ (or } x, y \in \mathbb{Z}), \quad (1.14)$$

are quasi-invariant, under the action of the group of compactly supported diffeomorphisms on \mathbb{R} (or under the action of $S(\infty)$ on \mathbb{Z}). The equivalence between *reduced Palm measures of the same order* plays a central rôle in the proof, which proceeds by the method, further developed in this paper, of *regularized multiplicative functionals*.

- **Determinantal point processes associated with Hilbert spaces of holomorphic functions.**

Holroyd and Soo [HS13] have shown that the determinantal point process with the Bergman kernel on the unit disk has the property of *insertion tolerance*: its Palm measures are equivalent to itself. For the Ginibre point process on the complex plane, using its finite-dimensional approximations by orthogonal polynomial ensembles, Osada and Shirai [OS14] have shown that Palm measures of different orders are singular, while Palm measures of the same orders are equivalent and the Radon-Nikodym derivative is a regularized multiplicative functional. In [BQ], the method of *regularized multiplicative functionals* has been further elaborated for obtaining in a unified way, on one hand, the equivalence of reduced Palm measures of *the same order* of the determinantal point processes on the complex plane \mathbb{C} with correlation kernels given by the reproducing kernels of generalized Fock spaces on \mathbb{C} , and on the other hand, the equivalence of reduced Palm measures of *all orders* of the determinantal point processes on the open unit disk $\mathbb{D} \subset \mathbb{C}$ with correlation kernels given by the reproducing kernels of generalized Bergman spaces on \mathbb{D} . Specifically, the Radon-Nikodym derivative between the determinantal point process with the Bergman kernel on the unit disk and its Palm measures is computed explicitly as a regularized Blaschke product.

As a consequence, we also obtained the quasi-invariance property of these determinantal point processes, under the action of the group of compactly supported diffeomorphisms on the complex plane \mathbb{C} and on the open unit disk \mathbb{D} respectively.

- **Relations with rigidity of determinantal point processes.**

Recall that a point process on a Euclidean space \mathbb{R}^d is said to be rigid in the sense of Ghosh [Gho14] and Ghosh-Peres[GP], if for any bounded open subset

$B \subset \mathbb{R}^d$, such that the topological boundary ∂B is negligible with respect to the d -dimensional Lebesgue measure, the number of particles of this point process inside the subset B is almost surely determined by the configuration outside of the subset B . Olshanski's problem is closely related to this rigidity property of determinantal point processes. In particular, if a determinantal point process is rigid in the above sense, then its reduced Palm measures of *different orders* are almost surely singular, see [BQ, Prop. 8.1]. Note that for processes with J -Hermitian kernels we encounter a rather different notion of rigidity. In the case of discrete phase spaces, however, the new notion can be reduced to the old one, as we shall now demonstrate.

• **Discrete phase spaces and the particle-hole duality**

Analogues of our main results, Theorem 1.4 and Theorem 1.9, can be formulated and proved in similar way when the phase space \mathbb{R}^* is replaced by \mathbb{Z} or $\mathbb{Z}' = \frac{1}{2} + \mathbb{Z}$ or any other discrete subsets in \mathbb{R} . In particular, in the case where the phase space is a discrete subsets of \mathbb{R} , our results are related to previous works [Ols11, Buf14] by doing *particle-hole duality* as follows: Let \mathbb{P} be a determinantal point process, say on \mathbb{Z}' . Assume that \mathbb{P} has a Hermitian correlation kernel K . Define the particle-hole duality on $\mathbb{Z}'_- = \mathbb{Z}' \cap \mathbb{R}_-$ as a map $dual : \text{Conf}(\mathbb{Z}') \rightarrow \text{Conf}(\mathbb{Z}')$ given by

$$dual(X) := (X \cap \mathbb{Z}'_+) \sqcup (\mathbb{Z}'_- \setminus X).$$

Then this particle-hole duality transform \mathbb{P} to a new point process $dual_*(\mathbb{P})$, which is again a determinantal point process. A correlation kernel of this new point process can be provided by

$$K^\circ := \text{sgn} \cdot K + \chi_{\mathbb{Z}'_-}.$$

Note that K° is J -Hermitian with respect to the partition $\mathbb{Z}' = \mathbb{Z}'_+ \sqcup \mathbb{Z}'_-$ and the orthogonal decomposition $\ell^2(\mathbb{Z}') = \ell^2(\mathbb{Z}'_+) \oplus \ell^2(\mathbb{Z}'_-)$. In general, the particle-hole duality transforms a rigid point process (see Definition 3.1) to a *balanced rigid* one and vice-versa. It transforms a quasi-invariant point process to a point process having balanced Palm equivalence property and vice-versa. In terms of correlation kernels, the particle-hole duality transforms Hermitian kernels to J -Hermitian ones and vice-versa.

At the same time, we would like to note that the particle-hole duality argument only works in the case where the phase spaces are discrete. This can be already seen on the level of correlation kernels, indeed, the kernel \widehat{K} defined in (1.4) corresponds to K° as above. Observe that \widehat{K} can not be used to define (extended) Fredholm determinants, and it is not the correlation kernel of any determinantal point process. Thus when the phase space is \mathbb{R}^* , processes with J -Hermitian kernels can not be transformed to processes with Hermitian kernels.

1.5 Organization of the paper and schemes of proofs

The paper is organized as follows.

In the preliminary part, §2, we briefly recall the definition of determinantal point processes and theory of reduced Palm measures. In particular, we collect the necessary results from [Lyt13] on the general determinantal point processes with J -Hermitian correlation kernels. Some standard properties of extended Fredholm determinants are also collected in §2. The proofs for these properties are postponed to the appendix in the end of the paper.

The main body of the paper is separated into two parts. The first part, §3, is devoted to the proofs of Theorem 1.4 and Theorem A; the second part, §4, is devoted to the proofs of Theorem 1.9 and Theorem B. These two parts are essentially independent from each other.

First part §3: proofs of Theorem 1.4 and Theorem A. Here we follow the scheme of Ghosh and Peres [GP]. Let \mathbb{P}_K be the determinantal point process as in Theorem 1.4. Our main task is to construct, after fixing an arbitrarily large interval $U_R = [-R, R]$, a sequence of *compactly supported* continuous functions $(\varphi_n)_{n \geq 1}$ defined on \mathbb{R} , such that $\varphi_n(x)$ tends to 1 uniformly on U_R when n tends to infinity. Moreover, the following limit relation holds:

$$\mathrm{Var}_{\mathbb{P}_K} \left(\sum_{x \in X} \mathrm{sgn}(x) \varphi_n(x) \right) \xrightarrow{n \rightarrow \infty} 0.$$

See also [BDQ, BQ, Buf15] for the use of the same method in the Hermitian case.

Second part §4: proofs of Theorem 1.9 and Theorem B. There are three main ingredients in the proofs of our main results in this part:

- (i) The J -Hermitian kernels for the determinantal point processes under consideration have L -kernels, that is, the determinantal point processes are L -processes;
- (ii) Under certain assumptions on the L -kernel of the initial determinantal point process μ_L , all the reduced Palm measures conditioned at an equal number of positions at both sides of $\mathbb{R}^* = \mathbb{R}_+ \sqcup \mathbb{R}_-$ are again L -processes, and the L -kernels for these reduced Palm measures have the form fLf , where f is certain bounded measurable function defined on \mathbb{R}^* .
- (iii) Under suitable assumptions on the kernel L and f , the two determinantal point processes μ_{fLf} and μ_L are proved to be equivalent and the Radon-Nikodym derivative $d\mu_{fLf}/d\mu_L$ can be computed explicitly as a *regularized multiplicative functional*.

The verification of part (ii) will be given in §4.1. The proof relies heavily on the algebraic structures of the L -kernels, see Condition 4 in §4.1.

Let us now explain part (iii). We will first prove a preliminary and abstract version in Proposition 4.10: under a certain condition on the L -kernel, if $\text{supp}(f - 1) \subset \mathbb{R}^*$ has a positive distance from the origin, then μ_{fLf} and μ_L are equivalent and the Radon-Nikodym derivative is given by a normalized multiplicative functional:

$$\frac{d\mu_{fLf}}{d\mu_L}(X) = \frac{\prod_{x \in X} |f(x)|^2}{\mathbb{E}_{\mu_L} \prod_{x \in X} |f(x)|^2}. \quad (1.15)$$

But for f_p defined by (1.11), $\text{supp}(f_p - 1) = \mathbb{R}^*$, it does not have positive distance from the origin. Moreover, the usual multiplicative functional $\prod_{x \in X} |f_p(x)|^2$ on the right-hand side of (1.15) does not converge absolutely. For overcoming this difficulty, we are led to use a new version of *regularized multiplicative functionals*. One ingredient in the formalism of this new version of regularized multiplicative functionals is the use of the twisted linear statistics:

$$\sum_{x \in X} \text{sgn}(x) \varphi(x), \quad (\text{where } X \in \text{Conf}(\mathbb{R}^*)).$$

Extra efforts are also required in dealing with the extended version of Fredholm determinants. The reader is referred to [Buf14, BQ] for the use of another version of regularized multiplicative functionals in computing Radon-Nikodym derivatives between determinantal point processes whose correlation kernels are Hermitian.

2 Preliminaries

Let \mathcal{E} be a locally compact complete metrizable separable space. Assume that on \mathcal{E} is equipped with a positive σ -finite Borel measure μ . A *configuration* on \mathcal{E} is defined to be an $\mathbb{N} \cup \{0\}$ -valued Radon measure on \mathcal{E} ; in other words, a configuration on \mathcal{E} is a collection of *particles*, possibly with multiplicity, that admits no accumulation points in \mathcal{E} . Let $\text{Conf}(\mathcal{E})$ denote the space of all configurations on \mathcal{E} . With respect to the topology induced by the vague topology on the space of Radon measures on \mathcal{E} , the space $\text{Conf}(\mathcal{E})$ is itself a complete metrizable separable space. A *point process* on \mathcal{E} is by definition a Borel probability measure on $\text{Conf}(\mathcal{E})$. For further background on point processes, see, e.g., [DVJ08].

We now briefly recall the definition of *determinantal point processes*, see, e.g., [Mac75, Sos00, Lyo03]. Fix a Radon measure μ on \mathcal{E} . A determinantal point process on \mathcal{E} is determined by a correlation kernel K , that is, a certain two-variable complex-valued function $K(x, y)$ on $\mathcal{E} \times \mathcal{E}$. More precisely, if we denote the determinantal point process with a correlation kernel K by \mathbb{P}_K , then this measure \mathbb{P}_K is completely determined by the following: for any positive integer $k \geq 1$ and any disjoint bounded subsets D_1, \dots, D_k of

\mathcal{E} , we have

$$\int_{\text{Conf}(\mathcal{E})} \prod_{i=1}^k \#(X \cap D_i) \mathbb{P}_K(dX) = \int_{D_1 \times \cdots \times D_k} \det(K(x_i, x_j))_{1 \leq i, j \leq k} d\mu^{\otimes k}(x_1 \cdots x_k). \quad (2.16)$$

The equality (2.16) implies that if D_1, \dots, D_r are disjoint bounded subsets of \mathcal{E} and k_i are positive integers, $k = k_1 + \cdots + k_r$, then

$$\begin{aligned} & \int_{\text{Conf}(\mathcal{E})} \prod_{i=1}^r \#(X \cap D_i) (\#(X \cap D_i) - 1) \cdots (\#(X \cap D_i) - k_i + 1) \mathbb{P}_K(dX) \\ &= \int_{D_1^{k_1} \times \cdots \times D_r^{k_r}} \det(K(x_i, x_j))_{1 \leq i, j \leq k} d\mu(x_1) \cdots d\mu(x_k). \end{aligned}$$

See, e.g., [HKPV09, Remark 1.2.3]. By definition, determinantal measures are always supported on the subset of *simple* configurations, that is, configurations all of whose particles have multiplicity one.

In this paper, we are particularly interested in the determinantal point processes with J -Hermitian correlation kernels, see §1.3 and §2.2 for a brief introduction on J -Hermitian kernels. The reader is referred to [Lyt13] for the general theory of such point processes.

2.1 Palm measures of determinantal point processes

In what follows, by Palm measures of a point process, we always mean its *reduced* Palm measures. Let us briefly recall the definition of Palm measures of determinantal point processes. For further details on Palm measures of general point processes, the reader is referred to [Kal86, DVJ08]

Let \mathbb{P} be a point process on \mathcal{E} (later, we will focus on the case $\mathcal{E} = \mathbb{R}$ or \mathbb{R}^*). Assume that for any positive integer k , the point process \mathbb{P} admits the k -th correlation measure ρ_k on \mathcal{E}^k , that is, ρ_k is a positive measure on \mathcal{E}^k such that for any disjoint bounded subsets D_1, \dots, D_k of \mathcal{E} , the following identity

$$\int_{\text{Conf}(\mathcal{E})} \prod_{i=1}^k \#(X \cap D_i) \mathbb{P}(dX) = \int_{D_1 \times \cdots \times D_k} \rho_k(dx_1 \cdots dx_k) \quad (2.17)$$

holds. Then for ρ_k -almost every k -tuple $\mathbf{q} = (q_1, \dots, q_k) \in \mathcal{E}^k$ of *distinct* points in \mathcal{E} , one can define a point process $\mathbb{P}^{\mathbf{q}}$ on \mathcal{E} by the following disintegration formula: for any non-negative Borel test function $u : \text{Conf}(\mathcal{E}) \times \mathcal{E}^k \rightarrow \mathbb{R}$,

$$\int_{\text{Conf}(\mathcal{E})} \sum_{q_1, \dots, q_k \in X}^* u(X; \mathbf{q}) \mathbb{P}(dX) = \int_{\mathcal{E}^k} \rho_k(d\mathbf{q}) \int_{\text{Conf}(\mathcal{E})} u(X \cup \{q_1, \dots, q_k\}; \mathbf{q}) \mathbb{P}^{\mathbf{q}}(dX), \quad (2.18)$$

where \sum^* denotes the sum over all distinct points $q_1, \dots, q_k \in X$. The point process \mathbb{P}^q is called the Palm measure of \mathbb{P} conditioned at q_1, \dots, q_k .

In the above situation, if the k -th correlation measure ρ_k for the point process \mathbb{P} is absolutely continuous with respect to the product measure $\mu^{\otimes k}$ on \mathcal{E}^k , then the Radon-Nikodym derivative

$$f_k(x_1, \dots, x_k) := \frac{d\rho_k}{d\mu^{\otimes k}}(x_1, \dots, x_k)$$

is called the k -th correlation function of \mathbb{P} . In terms of correlation functions, the Palm measure \mathbb{P}^q can be described as follows: it is a point process on \mathcal{E} such that its n -th correlation function is given by

$$f_n^q(x_1, \dots, x_n) = f_{n+k}(q_1, \dots, q_k, x_1, \dots, x_n).$$

Informally, if \mathcal{X} is a random configuration on \mathcal{E} whose probability distribution is given by the point process \mathbb{P} , then \mathbb{P}^q is the conditional distribution of the random configuration $\mathcal{X} \setminus \{q_1, \dots, q_k\}$ conditioned to the event that all particles q_1, \dots, q_k are in the configuration \mathcal{X} .

A Theorem of Shirai and Takahashi [ST03] states that the Palm measures of a determinantal measure are again determinantal measures. Let us formulate this result more precisely. Assume now \mathbb{P} is a determinantal point process on \mathcal{E} induced by a correlation kernel K , that is, $\mathbb{P} = \mathbb{P}_K$. Let $q \in \mathcal{E}$ and assume that $K(q, q) > 0$. Set

$$K^q(x, y) = K(x, y) - \frac{K(x, q)K(q, y)}{K(q, q)}. \quad (2.19)$$

If $K(q, q) = 0$, we set $K^q = K$. More generally, if $\mathbf{q} = (q_1, \dots, q_k) \in \mathcal{E}^k$ is a k -tuple of distinct points in \mathcal{E} , then we define by iteration

$$K^{\mathbf{q}} = (\dots (K^{q_1})^{q_2} \dots)^{q_k}. \quad (2.20)$$

Observe that the order of the points q_1, q_2, \dots, q_k has no effect in the above iteration.

Theorem 2.1 (Shirai and Takahashi [ST03]). *Let $\mathbb{P} = \mathbb{P}_K$ be a determinantal point process on \mathcal{E} induced by a correlation kernel K . Let $k \in \mathbb{N}$ be a positive integer. Then for ρ_k -almost every k -tuple $\mathbf{q} \in \mathcal{E}^k$ of distinct points in \mathcal{E} , the Palm measure $\mathbb{P}_K^{\mathbf{q}}$ of \mathbb{P}_K conditioned at \mathbf{q} is again a determinantal point process on \mathcal{E} . Moreover, $\mathbb{P}_K^{\mathbf{q}}$ is induced by the kernel $K^{\mathbf{q}}$ defined in (2.20), that is, we have*

$$\mathbb{P}_K^{\mathbf{q}} = \mathbb{P}_{K^{\mathbf{q}}}.$$

Remark 2.1. Theorem 2.1 was proved by Shirai and Takahashi in [ST03] for determinantal point processes with Hermitian correlation kernels. This result was independently proved

by Lyons in [Lyo03] in the case where the phase space is a discrete countable set and the correlation kernel corresponds to a Hermitian orthogonal projection. The proof in [ST03] can be generalized word by word for determinantal point processes without requiring that the correlation kernels are Hermitian.

2.2 J -Hermitian kernels and extended Fredholm determinants

Recall that in §1.3, we have defined the J -Hermitian kernels on $\mathbb{R}^* = \mathbb{R}_+ \sqcup \mathbb{R}_-$ as follows: a kernel $K : \mathbb{R}^* \times \mathbb{R}^* \rightarrow \mathbb{C}$ is called a J -Hermitian kernel if it defines a bounded linear operator on $L^2(\mathbb{R})$ and if

$$K(x, y) = \operatorname{sgn}(x)\operatorname{sgn}(y)\overline{K(y, x)}, \quad x, y \in \mathbb{R}^*.$$

In Theorem 1.3, we recalled the Lytvynov's characterization of the correlation kernels of determinantal point process in J -Hermitian case in our particular situation with the phase space $\mathbb{R}^* = \mathbb{R}_+ \sqcup \mathbb{R}_-$. We shall need a slight reformulation of Theorem 1.3.

Remark 2.2. Note that the determinantal point process \mathbb{P}_K induced by the kernel K as in Theorem 1.3 accumulates at infinity (both $+\infty$ and $-\infty$), in this situation, we will say that the *singularity of the kernel K* is at infinity. The change of variables $x \mapsto 1/x$ on \mathbb{R}^* transforms \mathbb{P}_K to a new determinantal point process on \mathbb{R}^* induced by the new kernel $\frac{1}{|xy|}K(1/x, 1/y)$. This new determinantal point process has a single accumulation point at the origin $0 \in \mathbb{R}$ of the real line, and in this situation, we call that the above new kernel has a singularity at the origin. Now it is clear how to formulate a version of Theorem 1.3 when the kernel K has a singularity at the origin (and there is no singularity at infinity): we just need to replace the conditions on Δ_1, Δ_2 required in Theorem 1.3 by the following new condition:

Δ_1 and Δ_2 are two measurable subsets of \mathbb{R} both having positive distances from 0.

Note that in the case of singularity at origin, the two subsets $\Delta_1, \Delta_2 \subset \mathbb{R}^*$ can be unbounded.

Let $\mathcal{L}_1(L^2(\mathbb{R}))$ denote the space of trace-class operators on $L^2(\mathbb{R})$ and let $\mathcal{L}_2(L^2(\mathbb{R}))$ denote the space of Hilbert-Schmidt operators on $L^2(\mathbb{R})$. For further details on trace-class and Hilbert-Schmidt operators, the reader is referred to [Sim05]. Following [Lyt13], we denote by $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$ the space of all bounded linear operators on $L^2(\mathbb{R}) = L^2(\mathbb{R}_+) \oplus L^2(\mathbb{R}_-)$ such that when written in the following block forms

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix},$$

we have $a, d \in \mathcal{L}_1(L^2(\mathbb{R}))$ and $b, c \in \mathcal{L}_2(L^2(\mathbb{R}))$. Clearly,

$$\mathcal{L}_1(L^2(\mathbb{R})) \subset \mathcal{L}_{1|2}(L^2(\mathbb{R})) \subset \mathcal{L}_2(L^2(\mathbb{R})). \quad (2.21)$$

Let K be a bounded linear operator on $L^2(\mathbb{R})$. Then for any subset $\Delta \subset \mathbb{R}$, we denote K^Δ the compressed operator defined by

$$K^\Delta := \chi_\Delta K \chi_\Delta.$$

By [Lyt13, Prop. 12], if K satisfies all the conditions in Theorem 1.3, including the condition that $0 \leq \widehat{K} \leq 1$, then for any bounded subset $\Delta \subset \mathbb{R}$, we have

$$K^\Delta \in \mathcal{L}_{1|2}(L^2(\mathbb{R})). \quad (2.22)$$

Similarly, if K satisfies all the conditions of the origin-singularity version of Theorem 1.3 as explained in Remark 2.2, then for any measurable subset $\Delta \subset \mathbb{R}$ having a positive distance from the origin, the compressed operator K^Δ belongs to $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$.

The space $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$ is a Banach space equipped with a norm $\|\cdot\|_{\mathcal{L}_{1|2}}$ defined by the following formula

$$\left\| \begin{bmatrix} a & b \\ c & d \end{bmatrix} \right\|_{\mathcal{L}_{1|2}} := \|a\|_1 + \|d\|_1 + \|b\|_2 + \|c\|_2,$$

where $\|\cdot\|_1$ is the trace-class norm while $\|\cdot\|_2$ is the Hilbert-Schmidt norm. Observe that $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$ is not an ideal in the C^* -algebra $\mathcal{L}(L^2(\mathbb{R}))$ of all bounded linear operators on $L^2(\mathbb{R})$.

We collect a few standard facts needed in what follows; for the reader's convenience, we include their proofs in the Appendix.

Proposition 2.2. *Let A, B be two operators in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. We have*

$$\|AB\|_{\mathcal{L}_{1|2}} \leq 2\|A\|_{\mathcal{L}_{1|2}}\|B\|_{\mathcal{L}_{1|2}}.$$

More generally, if A_1, \dots, A_n are operators in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$, then

$$\|A_1 \cdots A_n\|_{\mathcal{L}_{1|2}} \leq 2^{n-1} \prod_{i=1}^n \|A_i\|_{\mathcal{L}_{1|2}}.$$

Proposition 2.3. *Let $f : \mathbb{R} \rightarrow \mathbb{C}$ be a bounded measurable function and let K be an operator in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. Then the operators fK and Kf are both in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. Moreover,*

$$\max(\|fK\|_{\mathcal{L}_{1|2}}, \|Kf\|_{\mathcal{L}_{1|2}}) \leq \|f\|_\infty \|K\|_{\mathcal{L}_{1|2}},$$

where $\|f\|_\infty$ means the L^∞ -norm of f .

Proposition 2.4. *Let A, B be two operators in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. Assume that $1 + A$ is invertible. Then the operators $(1 + A)^{-1}B$ and $B(1 + A)^{-1}$ both belong to the class $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$.*

Recall that the trace of an operator $A \in \mathcal{L}_1(L^2(\mathbb{R}))$ is given by

$$\mathrm{tr}(A) = \sum_{n=1}^{\infty} (Ae_n, e_n),$$

where $\{e_n\}_{n=1}^{\infty}$ is an orthonormal basis of $L^2(\mathbb{R})$.

Let m be a positive integer. Denote by $\wedge^m(L^2(\mathbb{R}))$ the m -th antisymmetric tensor power of the Hilbert space $L^2(\mathbb{R})$. For any $A \in \mathcal{L}(L^2(\mathbb{R}))$, denote by $\wedge^m(A)$ the unique bounded linear operator on $\wedge^m(L^2(\mathbb{R}))$ determined by

$$\wedge^m(A)(v_1 \wedge \cdots \wedge v_m) = (Av_1) \wedge \cdots \wedge (Av_m), \quad v_1, \dots, v_m \in L^2(\mathbb{R}).$$

Definition 2.1 (Fredholm determinant, Grothendieck [Gro56]). Let $A \in \mathcal{L}_1(L^2(\mathbb{R}))$, then the Fredholm determinant $\det(1 + A)$ is defined by

$$\det(1 + A) := \sum_{m=0}^{\infty} \mathrm{tr}(\wedge^m(A)).$$

In [BOO00], it is proven that the function $A \mapsto \det(1 + A)$ admits a unique extension to $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$ which is continuous in the topology of $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. We will use the same notation $\det(1 + A)$ for this extended Fredholm determinant when $A \in \mathcal{L}_{1|2}(L^2(\mathbb{R}))$.

Proposition 2.5. Let A, B be two operators in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$, then

$$\det((1 + A)(1 + B)) = \det(1 + A) \det(1 + B). \quad (2.23)$$

Proposition 2.6. Let $A \in \mathcal{L}_{1|2}(L^2(\mathbb{R}))$ and let $f : \mathbb{R} \rightarrow \mathbb{C}$ be a bounded measurable function. Then

$$\det(1 + fA) = \det(1 + Af). \quad (2.24)$$

We also need the following characterization of determinantal point processes with J -Hermitian correlation kernels in terms of multiplicative functionals.

Theorem 2.7 (E. Lytvynov[Lyt13]). Let K be a kernel as in Theorem 1.3. Then the determinantal point process \mathbb{P}_K is uniquely determined by the following property: for any compactly supported bounded measurable function $f : \mathbb{R} \rightarrow \mathbb{R}$, if $\Delta \subset \mathbb{R}$ is a bounded subset such that $\mathrm{supp}(f) \subset \Delta$, then we have

$$\int_{\mathrm{Conf}(\mathbb{R})} \prod_{x \in X} (1 + f(x)) \mathbb{P}_K(dX) = \det(1 + fK^\Delta).$$

3 Balanced rigidity

For any bounded Borel subset $B \subset \mathcal{E}$, let $\#_B : \text{Conf}(\mathcal{E}) \rightarrow \mathbb{N}$ be defined by

$$\#_B(X) := \#(B \cap X).$$

Fix a Borel subset $C \subset \mathcal{E}$, let

$$\mathcal{F}_C = \sigma(\{\#_B : B \subset C, B \text{ Borel}\})$$

be the smallest σ -algebra making all functions from $\{\#_B : B \subset C, B \text{ Borel}\}$ measurable. If \mathbb{P} is a point process on \mathcal{E} , then we denote $\mathcal{F}_C^\mathbb{P}$ for the \mathbb{P} -completion of \mathcal{F}_C .

Definition 3.1 (Ghosh [Gho14], Ghosh-Peres[GP]). A point process \mathbb{P} on \mathbb{R} is called *rigid* if for any bounded measurable subset $B \subset \mathbb{R}$, the random variable $\#_{B \cap \mathbb{R}}$ is $\mathcal{F}_{\mathbb{R} \setminus B}^\mathbb{P}$ -measurable.

Definition 3.2 (Singularity at infinity version). A point process \mathbb{P} on \mathbb{R}^* is called *balanced rigid* with respect to the partition $\mathbb{R}^* = \mathbb{R}_+ \sqcup \mathbb{R}_-$ if for any bounded measurable subset $B \subset \mathbb{R}^*$, the random variable

$$\#_{B \cap \mathbb{R}_+} - \#_{B \cap \mathbb{R}_-}$$

is $\mathcal{F}_{\mathbb{R}^* \setminus B}^\mathbb{P}$ -measurable.

3.1 A sufficient condition for balanced rigidity

This section is devoted to the proof of Theorem 1.4.

Assume that K is a J -Hermitian kernel on \mathbb{R} satisfying Conditions 1 and 2. The operators K and \widehat{K} have the following block forms with respect to the decomposition $L^2(\mathbb{R}) = L^2(\mathbb{R}_+, dx) \oplus L^2(\mathbb{R}_-, dx)$:

$$K = \begin{bmatrix} K_{++} & K_{+-} \\ K_{-+} & K_{--} \end{bmatrix} \text{ and } \widehat{K} = \begin{bmatrix} K_{++} & K_{+-} \\ -K_{-+} & 1_{\mathbb{R}_-} - K_{--} \end{bmatrix}, \quad (3.25)$$

where for instance $K_{+-} : L^2(\mathbb{R}_-, dx) \rightarrow L^2(\mathbb{R}_+, dx)$ stands for the operator $K_{+-} = P_+ K P_-$ and $1_{\mathbb{R}_-}$ stands for the identity operator on $L^2(\mathbb{R}_-, dx)$. Note that the operator K_{+-} admits the following integral kernel

$$K_{+-}(x, y) = \chi_{\mathbb{R}_+}(x) K(x, y) \chi_{\mathbb{R}_-}(y).$$

Recall that in Condition 1, we assume that the operator

$$\widehat{K} := \text{sgn} \cdot K + \chi_{\mathbb{R}_-} = P_+ K + P_- (1 - K)$$

defines an orthogonal projection on $L^2(\mathbb{R})$.

Lemma 3.1. *For any $x \in \mathbb{R}^*$, we have*

$$K(x, x) = \int_{\mathbb{R}} |K(x, y)|^2 dy. \quad (3.26)$$

Proof. By assumption, \widehat{K} is an orthogonal projection, hence $\widehat{K}^2 = \widehat{K}$. By substituting (3.25) into this identity and considering the diagonal blocks, we deduce that

$$\begin{cases} K_{++} = K_{++}^2 - K_{+-}K_{-+} \\ K_{--} = K_{--}^2 - K_{-+}K_{+-} \end{cases}.$$

The above first identity combined with (1.3) implies (3.26) for $x > 0$ while the second one combined with (1.3) implies (3.26) when $x < 0$. \square

Given a Borel function $\varphi : \mathbb{R} \rightarrow \mathbb{R}$, we define $\varphi^\circ : \mathbb{R} \rightarrow \mathbb{R}$ by

$$\varphi^\circ(x) := \operatorname{sgn}(x)\varphi(x). \quad (3.27)$$

By definition, the linear statistic $S[\varphi]$ corresponding to φ is the following function on $\operatorname{Conf}(\mathbb{R})$:

$$S[\varphi](X) := \sum_{x \in X} \varphi(x), \quad (3.28)$$

provided the right-hand side converges absolutely. For simplifying the notation, we set

$$T[\varphi] := S[\varphi^\circ]. \quad (3.29)$$

Recall that by Theorem 1.3, the kernel K satisfying Condition 1 induce a determinantal point process on \mathbb{R} , denoted by \mathbb{P}_K .

Lemma 3.2. *Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be a Borel function such that*

$$\int_{\mathbb{R}} f(x)^2 K(x, x) dx < \infty.$$

Then we have

$$\operatorname{Var}_{\mathbb{P}_K}(T[f]) = \frac{1}{2} \iint_{\mathbb{R}^2} |f(x) - f(y)|^2 |K(x, y)|^2 dx dy, \quad (3.30)$$

where $\operatorname{Var}_{\mathbb{P}_K}(T[f])$ stands for the variance of the random variable $T[f]$ defined on the probability space $(\operatorname{Conf}(\mathbb{R}), \mathbb{P}_K)$ equipped with the Borel σ -algebra.

Proof. By definition of correlation functions of determinantal point process, we have

$$\begin{aligned} \operatorname{Var}_{\mathbb{P}_K}(T[f]) &= \int_{\mathbb{R}} f^\circ(x)^2 K(x, x) dx - \iint_{\mathbb{R}^2} f^\circ(x) f^\circ(y) K(x, y) K(y, x) dx dy \\ &= \int_{\mathbb{R}} f(x)^2 K(x, x) dx - \iint_{\mathbb{R}^2} f(x) f(y) |K(x, y)|^2 dx dy. \end{aligned}$$

Substituting the formula (3.26) into the above identity, we get the desired formula (3.30). \square

Lemma 3.3. *Let K be a kernel satisfying Conditions 1 and 2. Then for any fixed $R > 0$, there exists a sequence $(\varphi_n)_{n \in \mathbb{N}}$ of real-valued Schwartz functions, such that $|\varphi_n(x)| \leq 1$ and*

$$\lim_{n \rightarrow \infty} \sup_{x \in [-R, R]} |\varphi_n(x) - 1| = 0 \text{ and } \lim_{n \rightarrow \infty} \text{Var}_{\mathbb{P}_K}(T[\varphi_n]) = 0.$$

Proof. It suffices to prove that given any positive integer $n \in \mathbb{N}$, we can construct a real-valued Schwartz function φ_n such that

$$|\varphi_n(x)| \leq 1, \quad \sup_{x \in [-R, R]} |\varphi_n(x) - 1| \leq 1/n \text{ and } \text{Var}_{\mathbb{P}_K}(T[\varphi_n]) \leq 1/n.$$

Let $M > 0$ be the number given in Condition 2. Fix a real number $N > 1$ which will be specified later. Given a real-valued Schwartz function f , denote

$$F(x, y) := \frac{1}{2} |f(x) - f(y)|^2 |K(x, y)|^2.$$

We define $I_i(f)$, $i = 1, 2, 3, 4$ as follows:

$$\begin{aligned} \text{Var}_{\mathbb{P}_K}(T[f]) &\leq \underbrace{\iint_{|x| \leq NM, |y| \leq NM} F}_{=: I_1(f)} + \underbrace{\iint_{|x| \leq M, |y| \geq NM} F}_{=: I_2(f)} + \\ &\quad + \underbrace{\iint_{|x| \geq NM, |y| \leq M} F}_{=: I_3(f)} + \underbrace{\iint_{|x| \geq M, |y| \geq M} F}_{=: I_4(f)}. \end{aligned} \quad (3.31)$$

Step 1: Control of I_2 and I_3 .

We claim that

$$\lim_{N \rightarrow \infty} \iint_{\{|x| \leq M, |y| \geq N \cdot M\}} |K(x, y)|^2 dx dy = 0. \quad (3.32)$$

Indeed, by Lemma 3.1 and condition (1.5), we have

$$\iint_{\{|x| \leq M, |y| \geq N \cdot M\}} |K(x, y)|^2 dx dy \leq \int_{|x| \leq M} K(x, x) dx < \infty.$$

Then the claim in (3.32) follows from above inequality and bounded convergence theorem. Now let us choose $N \geq R + 1$ large enough, such that

$$\iint_{\{|x| \leq M, |y| \geq NM\}} |K(x, y)|^2 dx dy = \iint_{\{|x| \geq NM, |y| \leq M\}} |K(x, y)|^2 dx dy \leq \frac{1}{40n}.$$

It follows that for any function f such that $|f| \leq 1$, we have

$$I_2(f) + I_3(f) \leq \frac{1}{10n}.$$

In what follows, we fix N chosen as above.

Step 2: Control of I_1 .

Note that N being fixed, the number NM is also fixed. For any function f , we have

$$\begin{aligned} I_1(f) &\leq \left(\sup_{|x| \leq NM, |y| \leq NM} |f(x) - f(y)|^2 \right) \iint_{|x| \leq NM, |y| \leq NM} |K(x, y)|^2 dx dy \\ &\leq 4 \left(\sup_{|x| \leq NM} |f(x) - 1|^2 \right) \int_{|x| \leq NM} |K(x, x)| dx. \end{aligned}$$

It follows that for any f such that

$$\sup_{|x| \leq NM} |f(x) - 1| \leq \min \left\{ n^{-1}, \left(20n \cdot \int_{|x| \leq NM} K(x, x) dx \right)^{-1/2} \right\},$$

we have

$$I_1(f) \leq \frac{1}{10n}.$$

For future use, let us denote

$$\delta_n := \min \left\{ n^{-1}, \left(20n \cdot \int_{|x| \leq NM} K(x, x) dx \right)^{-1/2} \right\}. \quad (3.33)$$

Step 3: Control of I_4 .

By (1.7), we may write

$$\begin{aligned} I_4(f) &\leq \iint_{\mathbb{R}^2} |f(x) - f(y)|^2 \Phi(x - y) dx dy \\ &= \iint_{\mathbb{R}^2} |f(x + t) - f(x)|^2 \Phi(t) dx dt \\ &= \iint_{\mathbb{R}^2} |\widehat{f}(\xi)|^2 |e^{i2\pi t\xi} - 1|^2 \Phi(t) d\xi dt \\ &= \int_{\mathbb{R}} |\widehat{f}(\xi)|^2 (2\widehat{\Phi}(0) - \widehat{\Phi}(\xi) - \widehat{\Phi}(-\xi)) d\xi, \end{aligned} \quad (3.34)$$

where \widehat{f} and $\widehat{\Phi}$ are the Fourier transforms of f and Φ respectively. Now we will apply a result from [Boa67], which says that for a positive integrable function Φ , condition (1.6) is equivalent to

$$\widehat{\Phi}(\zeta + \xi) + \widehat{\Phi}(\zeta - \xi) - 2\widehat{\Phi}(\zeta) = O(|\xi|), \text{ uniformly in } \zeta, \text{ as } |\xi| \rightarrow 0. \quad (3.35)$$

Take $\zeta = 0$ in (3.35) and note that $\widehat{\Phi}$ is bounded, we see that there exists a numerical constant C which only depends on Φ , such that

$$2\widehat{\Phi}(0) - \widehat{\Phi}(\xi) - \widehat{\Phi}(-\xi) \leq C|\xi|, \text{ for all } \xi \in \mathbb{R}.$$

Substitute this inequality into the estimate (3.34), we obtain

$$I_4(f) \leq C \int_{\mathbb{R}} |\xi| |\widehat{f}(\xi)|^2 d\xi.$$

Step 4: Construction of φ_n .

Recall the definition of δ_n in (3.33). Let $k \geq n$ be large enough such that for any $|t| \leq NMk^{-1}$, we have

$$|e^{i2\pi t} - 1| \leq \delta_n.$$

We claim that there exists a non-negative even function $\psi_n \in C_c^\infty(\mathbb{R})$ supported in a $(\frac{1}{k})$ -neighbourhood of 0, such that

$$\int_{\mathbb{R}} \psi_n(\xi) d\xi = 1 \text{ and } \int_{\mathbb{R}} |\xi| \psi_n(\xi)^2 d\xi \leq \frac{1}{10Cn}. \quad (3.36)$$

Indeed, since the function $\frac{1}{10Cn|\xi|} \chi_{|\xi| \leq 1/k}$ is not integrable, we can easily construct a Schwartz function ψ_n such that

$$\int_{\mathbb{R}} \psi_n = 1 \text{ and } \psi_n(\xi) \leq \frac{1}{10Cn|\xi|} \chi_{|\xi| \leq 1/k}, \text{ for any } \xi \in \mathbb{R}.$$

This last pointwise inequality implies that $\text{supp}(\psi_n) \subset [-1/k, 1/k]$ and

$$\int_{\mathbb{R}} |\xi| \psi_n(\xi)^2 d\xi \leq \left(\sup_{\xi} |\xi| \psi_n(\xi) \right) \cdot \int_{\mathbb{R}} \psi_n(\xi) d\xi \leq \frac{1}{10Cn}.$$

Now set

$$\varphi_n(x) = \check{\psi}_n(x) = \int_{\mathbb{R}} \psi_n(\xi) e^{i2\pi x \xi} d\xi.$$

Then $\varphi_n \in \mathcal{S}(\mathbb{R})$, $\varphi_n(0) = 1$ and $|\varphi_n(x)| \leq 1$. Moreover, since ψ_n is an even real-valued function, φ_n is real-valued. By construction, we have

$$I_4(\varphi_n) \leq C \int_{\mathbb{R}} |\xi| |\widehat{\varphi}_n(\xi)|^2 d\xi \leq \frac{1}{10n}.$$

Moreover, by our choice of k , we know that if $|\xi| \leq k^{-1}$ and $|x| \leq NM$, then we have $|e^{i2\pi x \xi} - 1| \leq \delta_n$. Hence for any $|x| \leq NM$,

$$\begin{aligned} |\varphi_n(x) - 1| &= |\varphi_n(x) - \varphi_n(0)| \leq \int_{\mathbb{R}} |e^{i2\pi x \xi} - 1| |\psi_n(\xi)| d\xi \\ &= \int_{|\xi| \leq k^{-1}} |e^{i2\pi x \xi} - 1| |\psi_n(\xi)| d\xi \leq \delta_n \leq n^{-1}. \end{aligned}$$

By Step 2, the above inequality implies that $I_1(\varphi_n) \leq \frac{1}{10n}$. It is readily seen that we also have $I_i(\varphi_n) \leq \frac{1}{10n}$, $i = 2, 3$, hence

$$\text{Var}_{\mathbb{P}_K}(T[\varphi_n]) \leq \sum_{i=1}^4 I_i(\varphi_n) \leq \frac{1}{n}.$$

This completes the proof of the proposition. \square

Remark 3.1. The construction in (3.36) relies heavily on the non-integrability of $\frac{1}{|\xi|}$ in any neighbourhood of the origin. Indeed, given a positive function $a(\xi)$,

$$\left(\int_{\mathbb{R}} a(\xi)^{-1} d\xi \right)^{-1} = \inf \left\{ \int_{\mathbb{R}} a(\xi) \psi(\xi)^2 d\xi : \psi \text{ positive and } \int_{\mathbb{R}} \psi(\xi) d\xi = 1 \right\},$$

with the understanding that the left hand side equals to 0 if $a(\xi)^{-1}$ is not integrable.

Now we can prove Theorem 1.4. Our proof follows the line of that of [GP, Thm. 6.1].

Proof of Theorem 1.4. Let $B \subset \mathbb{R}$ be any bounded measurable subset. Choose $R > 0$ large enough such that $B \subset [-R, R]$. Let φ_n be a sequence of Schwartz functions constructed as in Lemma 3.3. We have

$$T[\varphi_n](X) = \sum_{x \in X \cap B} \varphi_n(x) \operatorname{sgn}(x) + \sum_{x \in X \setminus B} \varphi_n(x) \operatorname{sgn}(x) =: I(n) + II(n).$$

First note that

$$\begin{aligned} \|I(n) - \sum_{x \in X \cap B} \operatorname{sgn}(x)\|_1 &\leq \mathbb{E}_{\mathbb{P}_K} \sum_{x \in X} |1 - \varphi_n(x)| \chi_B(x) \\ &\leq \sup_{x \in [-R, R]} |1 - \varphi_n(x)| \cdot \int_B K(x, x) dx, \end{aligned}$$

we have, passing to a subsequence if necessary,

$$I(n) \xrightarrow[\mathbb{P}_K\text{-almost surely}]{n \rightarrow \infty} \#_{B \cap \mathbb{R}_+} - \#_{B \cap \mathbb{R}_-}. \quad (3.37)$$

By construction, $\lim_{n \rightarrow \infty} \operatorname{Var}_{\mathbb{P}_K}(T[\varphi_n]) = 0$, passing to a subsequence if necessary, we have

$$I(n) + II(n) - \mathbb{E}_{\mathbb{P}_K} T[\varphi_n] \xrightarrow[\mathbb{P}_K\text{-almost surely}]{n \rightarrow \infty} 0. \quad (3.38)$$

Combining (3.37) and (3.38), for \mathbb{P}_K -almost every configuration $X \in \operatorname{Conf}(\mathbb{R})$, we get

$$\#_{B \cap \mathbb{R}_+}(X) - \#_{B \cap \mathbb{R}_-}(X) = \lim_{n \rightarrow \infty} (\mathbb{E}_{\mathbb{P}_K} T[\varphi_n] - II(n)(X)).$$

Since all the functions $\mathbb{E}_{\mathbb{P}_K} T[\varphi_n] - II(n)$ are $\mathcal{F}_{\mathbb{R} \setminus B}$ -measurable, the \mathbb{P}_K -almost sure limit function $\#_{B \cap \mathbb{R}_+} - \#_{B \cap \mathbb{R}_-}$ is $\mathcal{F}_{\mathbb{R} \setminus B}^{\mathbb{P}_K}$ -measurable. The proof of Theorem 1.4 is complete. \square

3.2 Proof of Theorem A

Following [BO00, Thm. 5.3], when z, z' are fixed, we denote the Whittaker kernel $\mathcal{K}_{z, z'}$ simply by \mathcal{K} . The change of variables $x \mapsto 1/x$ transforms the Whittaker kernel to the following new kernel

$$K_{\text{new}}(x, y) = \frac{1}{|xy|} \mathcal{K}(1/x, 1/y). \quad (3.39)$$

Note that the kernel K_{new} satisfies Condition 1. Indeed, it is known in [BO00] that the Whittaker kernel \mathcal{K} is such that the operators $P_+\mathcal{K}P_+$ and $P_-\mathcal{K}P_-$ are non-negative and for any subsets Δ_1, Δ_2 both with positive distance from the origin such that $\Delta_1 \subset \mathbb{R}_+$ and $\Delta_2 \subset \mathbb{R}_-$, the operators $\chi_{\Delta_i}\mathcal{K}\chi_{\Delta_i}$ ($i = 1, 2$) are in trace-class, and the operator $\chi_{\Delta_2}\mathcal{K}\chi_{\Delta_1}$ is Hilbert-Schmidt. Moreover the operator

$$\widehat{\mathcal{K}} := \text{sgn} \cdot \mathcal{K} + \chi_{\mathbb{R}_-}$$

defines an *orthogonal projection* on $L^2(\mathbb{R}, dx)$. By the change of variable $x \mapsto 1/x$, these properties imply exactly that the kernel K_{new} satisfies Condition 1.

Lemma 3.4. *Assume that the parameters z, z' satisfy the conditions $z' = \bar{z}$ and $z \in \mathbb{C} \setminus \mathbb{R}$. Then K_{new} satisfies Condition 2.*

Proof. From the explicit formula (1.2) for the Whittaker kernel, we see that the diagonal value $\mathcal{K}(x, x)$ is given by

$$\mathcal{K}(x, x) = \pm(\mathcal{P}'_{\pm}(|x|)\mathcal{Q}_{\pm}(|x|) - \mathcal{Q}'_{\pm}(|x|)\mathcal{P}_{\pm}(|x|)), \quad (3.40)$$

the sign \pm depends on the sign $\text{sgn}(x)$ of the real number $x \in \mathbb{R}^*$. Since the Whittaker function converges to 0 exponentially fast at infinity, it is readily seen that

$$\int_{|x|>\delta} \mathcal{K}(x, x)dx < \infty. \quad (3.41)$$

This in turn implies the condition (1.5) for K_{new} around the origin.

As in the proof of [Bor, Prop. 4.1.3], for $x > 0$ near the origin, by expressing the Whittaker functions in terms of confluent hypergeometric functions, the functions \mathcal{P}_+ and \mathcal{Q}_+ can be written as

$$\begin{aligned} \mathcal{P}_+(x) &= x^{\frac{z-z'}{2}} A_1(x) + x^{\frac{z'-z}{2}} B_1(x) \\ \mathcal{Q}_+(x) &= x^{\frac{z-z'}{2}} A_2(x) + x^{\frac{z'-z}{2}} B_2(x) \end{aligned} \quad (3.42)$$

where $A_i(x), B_i(x), i = 1, 2$, are analytic in a neighbourhood of the origin. At infinity, both functions tend to 0. Hence \mathcal{P}_+ and \mathcal{Q}_+ are bounded on \mathbb{R}_+ . The fact that \mathcal{P}_- and \mathcal{Q}_- are bounded on \mathbb{R}_- can be proved similarly. It follows that there exists $C > 0$, such that

$$\text{for any } x, y \in \mathbb{R}^*, |\mathcal{K}(x, y)| \leq \frac{C}{|x - y|},$$

or, equivalently,

$$\text{for any } x, y \in \mathbb{R}, |K_{\text{new}}(x, y)| \leq \frac{C}{|x - y|}. \quad (3.43)$$

Now fix $M > 0$, let δ' be a fixed number such that $0 < \delta' < M/4$. We claim that there exists $C' > 0$, such that

$$\text{if } |x| \geq M, |y| \geq M \text{ and } |x - y| \leq \delta', \text{ then } |K_{\text{new}}(x, y)| \leq C'. \quad (3.44)$$

Indeed, by the choice of δ' , any pair (x, y) verifying the hypothesis in (3.44) satisfies $\text{sgn}(x)\text{sgn}(y) > 0$. If $x > 0, y > 0$, then

$$\begin{aligned} K_{\text{new}}(x, y) &= \frac{\mathcal{P}_+(\frac{1}{x})\mathcal{Q}_+(\frac{1}{y}) - \mathcal{Q}_+(\frac{1}{x})\mathcal{P}_+(\frac{1}{y})}{y - x} \\ &= -\mathcal{P}_+\left(\frac{1}{x}\right)\mathcal{Q}'_+\left(\frac{1}{\xi_{x,y}}\right)\frac{1}{\xi_{x,y}^2} + \mathcal{Q}_+\left(\frac{1}{x}\right)\mathcal{P}'_+\left(\frac{1}{\xi_{x,y}}\right)\frac{1}{\xi_{x,y}^2}, \end{aligned} \quad (3.45)$$

where $\xi_{x,y} \in (\min(x, y), \max(x, y))$. By (3.42), it is readily seen that

$$\mathcal{P}'_+(1/x) = O(x), \quad \mathcal{Q}'_+(1/x) = O(x), \quad \text{as } x \rightarrow \infty. \quad (3.46)$$

From (3.45) and (3.46), it is readily seen that (3.44) holds for $x > 0, y > 0$. Similarly, by analyzing $\mathcal{P}_-, \mathcal{Q}_-$, we also obtain (3.44) for $x < 0, y < 0$. Combining (3.43) and (3.44), we see that the condition (1.7) in Condition 2 holds for K_{new} , that is

$$\text{if } |x| \geq M, |y| \geq M, \text{ then } |K_{\text{new}}(x, y)|^2 \leq \Phi(x - y),$$

where

$$\Phi(t) = (C')^2 \chi_{|t| \leq \delta'} + \frac{C^2}{t^2} \chi_{|t| \geq \delta'},$$

is a function satisfying the required condition (1.6). \square

4 Balanced Palm equivalence property

Recall that for a point process \mathbb{P} on \mathcal{E} and a positive integer $k \in \mathbb{N}$, the k -th correlation measure ρ_k of \mathbb{P} is a positive measure on \mathcal{E}^k , which is defined by the relation (2.17).

Definition 4.1. A point process \mathbb{P} on \mathcal{E} is said to have *balanced Palm equivalence property* with respect to the partition $\mathcal{E} = \mathcal{E}_1 \sqcup \mathcal{E}_2$, if for any positive integer $n \in \mathbb{N}$, for ρ_{2n} -almost every $2n$ -tuple $\mathbf{p} \in \mathcal{E}_1^n \times \mathcal{E}_2^n$ of distinct points, in other words, \mathbf{p} is a $2n$ -tuple of distinct points of \mathcal{E} with a equal number of points from \mathcal{E}_1 and \mathcal{E}_2 , the Palm measure $\mathbb{P}^{\mathbf{p}}$ is equivalent to \mathbb{P} .

For processes governed by J -Hermitian kernels, the balanced Palm equivalence property is the natural analogue of equivalence of Palm measures of the same order for processes with Hermitian kernels.

4.1 Palm measures of L -processes

In this section, we will study the correlation kernels of Palm measures for L -processes.

Recall the definition of L -kernels in Definition 1.1. Let μ_L be the determinantal measure induced by a kernel $K_L = L(1+L)^{-1}$, where L is a kernel satisfying Condition 3. By Lemma 1.6, the kernel K_L is J -Hermitian. When L is fixed, we simply write $K = K_L$. Shirai-Takahashi's Theorem 2.1 says that for almost every $p \in \mathbb{R}^*$ (with respect to the measure $K(x, x)dx$), the Palm measure μ_L^p is a determinantal point process with the following kernel:

$$K^p(x, y) = K(x, y) - \frac{K(x, p)K(p, y)}{K(p, p)} = K(x, y) - \operatorname{sgn}(p)\operatorname{sgn}(y) \frac{K(x, p)\overline{K(y, p)}}{K(p, p)}. \quad (4.47)$$

Let $p^+ > 0$ and $p^- < 0$, our aim is to describe the correlation kernel of the Palm measure $\mu_L^{(p^+, p^-)}$, that is, the kernel

$$K^{(p^+, p^-)} := (K^{p^+})^{p^-} = (K^{p^-})^{p^+}.$$

More generally, we are going to describe the kernel $K^{\mathbf{p}}$ defined by the formula (2.20) when $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-)$ with $p_i^+ > 0$ and $p_i^- < 0$ for $i = 1, \dots, n$. While it is easily seen that $K^{\mathbf{p}}$ is a J -Hermitian kernel satisfying Condition 1 (transformed to the version with singularity at origin), it is a priori not clear whether $K^{\mathbf{p}}$ admits an L -kernel. We now check that it does and that the L -kernel of $K^{\mathbf{p}}$ also satisfies Condition 3.

Definition 4.2. Given $\mathbf{p} = (p^+, p^-)$, where $p^+ > 0, p^- < 0$, we define a *bounded* function on \mathbb{R}^* by the formula

$$g_{\mathbf{p}}(x) = \frac{x - p^+}{x - p^-} \chi_{\{x > 0\}} + \frac{x - p^-}{x - p^+} \chi_{\{x < 0\}}. \quad (4.48)$$

More generally, if $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-)$ with $p_i^+ > 0$ and $p_i^- < 0$ for $i = 1, \dots, n$, we set

$$g_{\mathbf{p}}(x) = \prod_{i=1}^n \left(\frac{x - p_i^+}{x - p_i^-} \chi_{\{x > 0\}} + \frac{x - p_i^-}{x - p_i^+} \chi_{\{x < 0\}} \right). \quad (4.49)$$

Proposition 4.1. Let L be an operator satisfying Condition 3. If $\mathbf{p} = (p^+, p^-)$ such that $p^+ > 0, p^- < 0$, then we have

$$K_L^{\mathbf{p}} = K_{g_{\mathbf{p}} L g_{\mathbf{p}}}.$$

Corollary 4.2. Let L be an operator satisfying Condition 3. Let

$$\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-)$$

be a $2n$ -tuple of real numbers such that $p_i^+ > 0$ and $p_i^- < 0$ for $i = 1, \dots, n$, then

$$K_L^{\mathbf{p}} = K_{g_{\mathbf{p}} L g_{\mathbf{p}}}.$$

Proof. When $\mathbf{p} = (p_1^+, p_1^-)$, this is just Proposition 4.1. Now since the new kernel

$$g_{(p_1^+, p_1^-)}(x)L(x, y)g_{(p_1^+, p_1^-)}(y)$$

has a similar structure as $L(x, y)$, that is, it satisfies Condition 3, we can continue our procedure and complete the proof of the corollary. \square

Proof of Proposition 1.7. Note that in Corollary 4.2, we obtain $K_L^{\mathbf{p}} = K_{g_{\mathbf{p}}Lg_{\mathbf{p}}}$. However, by the special form of L , we have (see Lemma 4.11 below for this fact)

$$g_{\mathbf{p}}Lg_{\mathbf{p}} = f_{\mathbf{p}}Lf_{\mathbf{p}},$$

where $g_{\mathbf{p}}$ and $f_{\mathbf{p}}$ are functions defined in (4.49) and (1.11) respectively. Hence we obtain that

$$\mu_L^{\mathbf{p}} = \mu_{f_{\mathbf{p}}Lf_{\mathbf{p}}}.$$

\square

Lemma 4.3. *Let $p \in \mathbb{R}^*$. Then the kernel $K^p(x, y)$ defined in (4.47) is J -Hermitian and \widehat{K}^p is an orthogonal projection. Moreover, if $p^+ > 0$, then*

$$\text{Ran}(\widehat{K}^{p^+}) = \text{Ran}(\widehat{K}) \ominus \mathbb{C}\text{sgn}(\cdot)K(\cdot, p^+);$$

if $p^- < 0$, then

$$\text{Ran}(\widehat{K}^{p^-}) = \text{Ran}(\widehat{K}) \oplus \mathbb{C}\text{sgn}(\cdot)K(\cdot, p^-).$$

Proof. It is clear that the kernel $K^p(x, y)$ is J -Hermitian. By Lemma 3.1, we see that, for any $p \in \mathbb{R}^*$, the following kernel

$$\ell_p(x, y) = \text{sgn}(x)\text{sgn}(y)\frac{K(x, p)\overline{K(y, p)}}{K(p, p)}$$

induces the orthogonal projection onto the one dimensional subspace $\mathbb{C}\text{sgn}(\cdot)K(\cdot, p)$. Let us denote this one dimensional projection again by ℓ_p . By definition, it is easy to see that

$$\widehat{K}^p = \widehat{K} - \text{sgn}(p)\ell_p.$$

That is, if $p^+ > 0$, then $\widehat{K}^{p^+} = \widehat{K} - \ell_{p^+}$ and if $p^- < 0$, then $\widehat{K}^{p^-} = \widehat{K} + \ell_{p^-}$. Thus for proving Lemma 4.3, we only need to show that

$$\text{sgn}(\cdot)K(\cdot, p^+) \in \text{Ran}(\widehat{K}) \text{ and } \text{sgn}(\cdot)K(\cdot, p^-) \in \text{Ran}(\widehat{K})^\perp. \quad (4.50)$$

The first relation in (4.50) is equivalent to

$$\int_{\mathbb{R}} K(x, y)\text{sgn}(y)K(y, p^+)dy + \chi_{\mathbb{R}^-}(x)K(x, p^+) = K(x, p^+); \quad (4.51)$$

while the second is equivalent to

$$\int_{\mathbb{R}} K(x, y) \operatorname{sgn}(y) K(y, p^-) dy + \chi_{\mathbb{R}_-}(x) K(x, p^-) = 0. \quad (4.52)$$

By using the fact that $\widehat{K}^2 = \widehat{K}$ and comparing all the block coefficients of the operator \widehat{K}^2 and \widehat{K} , both written in the block form as in (3.25), we get

$$\begin{cases} K_{++} = K_{++}^2 - K_{+-}K_{-+} \\ K_{--} = K_{--}^2 - K_{-+}K_{+-} \\ K_{++}K_{+-} = K_{+-}K_{--} \\ K_{-+}K_{++} = K_{--}K_{-+} \end{cases}.$$

The above first identity implies (4.51) for $x > 0$; the second one implies (4.52) for $x < 0$; the third one implies (4.52) for $x > 0$ and the last one implies (4.51) for $x < 0$. \square

Remark 4.1. Although \widehat{K} is not the reproducing kernel of the Hilbert subspace $\operatorname{Ran}(\widehat{K})$, the space $\operatorname{Ran}(\widehat{K})$ still possesses certain reproducing feature. Indeed, if $\varphi \in \operatorname{Ran}(\widehat{K})$, then we have the following identity of functions in $L^2(\mathbb{R})$:

$$\begin{aligned} \chi_{\mathbb{R}_+}(x) \varphi(x) &= \int_{\mathbb{R}} K(x, y) \varphi(y) dy = \int_{\mathbb{R}} \operatorname{sgn}(y) \overline{K(y, x)} \varphi(y) dy \\ &= \langle \varphi, \operatorname{sgn}(\cdot) K(\cdot, x) \rangle_{L^2(\mathbb{R})}. \end{aligned}$$

Now we can apply Lemma 4.3 to K^{p^+} and K^{p^-} respectively and get the following

Proposition 4.4. *Let $\mathbf{p} = (p^+, p^-)$ with $p^+ > 0, p^- < 0$, then*

$$\operatorname{Ran}(\widehat{K}^{\mathbf{p}}) = \left(\operatorname{Ran}(\widehat{K}) \ominus \mathbb{C} \operatorname{sgn}(\cdot) K(\cdot, p^+) \right) \oplus \mathbb{C} \operatorname{sgn}(\cdot) K^{p^+}(\cdot, p^-) \quad (4.53)$$

$$= \left(\operatorname{Ran}(\widehat{K}) \oplus \mathbb{C} \operatorname{sgn}(\cdot) K(\cdot, p^-) \right) \ominus \mathbb{C} \operatorname{sgn}(\cdot) K^{p^-}(\cdot, p^+). \quad (4.54)$$

We also need an explicit description of subspaces as $\operatorname{Ran}(\widehat{K}_L)$. It is convenient for us to introduce a general condition on the kernel L .

Condition 4. The L -operator is assumed to have the following block form with respect to the decomposition $L^2(\mathbb{R}^*, dx) = L^2(\mathbb{R}_+, dx) \oplus L^2(\mathbb{R}_-, dx)$:

$$L = \begin{bmatrix} 0 & V \\ -V^* & 0 \end{bmatrix}, \quad (4.55)$$

where $V : L^2(\mathbb{R}_-) \rightarrow L^2(\mathbb{R}_+)$ is a bounded linear operator. Moreover, assume that the operator V is such that for any $\varepsilon > 0$, the operators $\chi_{(\varepsilon, \infty)} V$ and $V \chi_{(-\infty, -\varepsilon)}$ are Hilbert-Schmidt.

L -kernels satisfying Condition 4 appear naturally in many contexts, see e.g. [BOO00].

Proposition 4.5 ([BO05], Prop. 5. 1). *Let L be an operator as in Condition 4, then the operator \widehat{K}_L is an orthogonal projection, the range $\text{Ran}(\widehat{K}_L)$ and its orthogonal complement are given by*

$$\text{Ran}(\widehat{K}_L) = \left\{ Vh \oplus h : h \in L^2(\mathbb{R}_-, dx) \right\}; \quad (4.56)$$

$$\text{Ran}(\widehat{K}_L)^\perp = \left\{ f \oplus (-V^*f) : f \in L^2(\mathbb{R}_+, dx) \right\}. \quad (4.57)$$

Remark 4.2. Let $f : \mathbb{R}^* \rightarrow \mathbb{C}$ be a bounded function, then

$$fLf = \begin{bmatrix} 0 & f\chi_{\mathbb{R}_+}Vf\chi_{\mathbb{R}_-} \\ -f\chi_{\mathbb{R}_-}V^*f\chi_{\mathbb{R}_+} & 0 \end{bmatrix} = \begin{bmatrix} 0 & f^+Vf^- \\ -f^-V^*f^+ & 0 \end{bmatrix}.$$

Proof of Proposition 4.1. Since $K \leftrightarrow \widehat{K}$ is a bijection, to show $K_L^p = K_{g_p L g_p}$ is equivalent to show the coincidence of two orthogonal projections:

$$\widehat{K}_L^p = \widehat{K_{g_p L g_p}}. \quad (4.58)$$

By Proposition 4.5 and Remark 4.2, we have

$$\text{Ran}(\widehat{K_{g_p L g_p}}) = \left\{ g_p^+ V(g_p^- h) \oplus h : h \in L^2(\mathbb{R}_-, dx) \right\}.$$

Hence to show the identity (4.58), it suffices to show the coincidence of the following two subspaces

$$\text{Ran}(\widehat{K}_L^p) = \left(\text{Ran}(\widehat{K}_L) \ominus \mathbb{C} \text{sgn}(\cdot) K_L(\cdot, p^+) \right) \oplus \mathbb{C} \text{sgn}(\cdot) K_L^{p^+}(\cdot, p^-)$$

and

$$\text{Ran}(\widehat{K_{g_p L g_p}}) = \left\{ g_p^+ V(g_p^- h) \oplus h : h \in L^2(\mathbb{R}_-, dx) \right\}. \quad (4.59)$$

Step 1: If $\varphi \in \text{Ran}(\widehat{K}_L) \ominus \mathbb{C} \text{sgn}(\cdot) K_L(\cdot, p^+)$, then $\varphi \in \text{Ran}(\widehat{K_{g_p L g_p}})$.

Recall that

$$\text{Ran}(\widehat{K}_L) = \left\{ Vh \oplus h : h \in L^2(\mathbb{R}_-, dx) \right\}.$$

Hence the hypothesis $\varphi \in \text{Ran}(\widehat{K}_L) \ominus \mathbb{C} \text{sgn}(\cdot) K_L(\cdot, p^+)$ is equivalent to the existence of a function $h \in L^2(\mathbb{R}_-)$ such that

$$\varphi = h + V(h) \text{ and } \varphi \perp \text{sgn}(\cdot) K_L(\cdot, p^+).$$

By Remark 4.1, this last condition can be translated to the condition $V(h)(p^+) = 0$, that is

$$A(p^+) \int_{\mathbb{R}_-} \frac{A(y)h(y)}{p^+ - y} dy = 0.$$

Since A is assumed to have full support, the set $\{p \in \mathbb{R}^* : A(p) = 0\}$ is negligible, hence we may assume that $A(p^+) \neq 0$. Thus we have

$$\int_{\mathbb{R}_-} \frac{A(y)h(y)}{p^+ - y} dy = 0.$$

Now we want to show that there exists $h_1 \in L^2(\mathbb{R}_-)$ such that

$$h + V(h) = h_1 + g_p^+ V(h_1 g_p^-).$$

The above identity is equivalent to

$$h = h_1 \text{ and } V(h) = g_p^+ V(h_1 g_p^-).$$

Hence what we need to show is: once we have $V(h)(p^+) = 0$, then

$$V(h) = \frac{x - p^+}{x - p^-} V\left(\frac{x - p^-}{x - p^+} h\right).$$

The above assertion is equivalent to

$$\frac{1}{x - p^+} V(h)(x) = V\left(\frac{1}{x - p^+} h\right)(x). \quad (4.60)$$

If we denote $k = \frac{h}{x - p^+}$, then the identity (4.60) is equivalent to

$$[x, V]k = 0, \quad (4.61)$$

where $[x, V]$ is the commutator between the multiplication x and V . Since the commutator $[x, V]$ has a kernel given by the formula $\chi_{\mathbb{R}_+}(x)A(x)A(y)\chi_{\mathbb{R}_-}(y)$, hence the identity (4.61) can be checked as follows:

$$([x, V]k)(x) = \chi_{\mathbb{R}_+}(x)A(x) \int_{\mathbb{R}_-} A(y)k(y)dy = \chi_{\mathbb{R}_+}(x)A(x) \int_{\mathbb{R}_-} \frac{A(y)h(y)}{y - p^+} dy = 0.$$

Step 2: If $\varphi = \text{sgn}(\cdot)K_L^{p^+}(\cdot, p^-)$, then $\varphi \in \text{Ran}(\widehat{K_{g_p L g_p}})$.

By (4.59), what we need to show is that

$$\varphi(x)\chi_{\mathbb{R}_+}(x) = \frac{x - p^+}{x - p^-} V\left(\frac{x - p^-}{x - p^+} \varphi(x)\chi_{\mathbb{R}_-}(x)\right)(x). \quad (4.62)$$

This is in turn equivalent to the following assertion: for $x > 0$, we have

$$K_L^{p^+}(x, p^-) = -\frac{x - p^+}{x - p^-} V\left(\frac{x - p^-}{x - p^+} K_L^{p^+}(\cdot, p^-)\right)(x). \quad (4.63)$$

By a result in [IKS90, Section II], under Condition 3, the kernel $K(x, y) = K_L(x, y)$ has the following integrable form

$$K(x, y) = \frac{F_1(x)G_1(y) + F_2(x)G_2(y)}{x - y},$$

where

$$\begin{cases} (1 + L)F_1 = A^+ \\ (1 + L)F_2 = A^- \\ (1 + L^*)G_1 = A^- \\ (1 + L^*)G_2 = A^+ \end{cases}.$$

Note that $L = V - V^*$, and since V, V^* has range in $L^2(\mathbb{R}_+), L^2(\mathbb{R}_-)$ respectively, the above equation system is equivalent to

$$\begin{cases} (F_1)_- - V^*F_1 = 0, & (F_1)_+ + VF_1 = A^+ \\ (F_2)_- - V^*F_2 = A^-, & (F_2)_+ + VF_2 = 0 \\ (G_1)_- + V^*G_1 = A^-, & (G_1)_+ - VG_1 = 0 \\ (G_2)_- + V^*G_2 = 0, & (G_2)_+ - VG_2 = A^+ \end{cases}. \quad (4.64)$$

Moreover, we have

$$F_1(x)G_1(x) + F_2(x)G_2(x) = 0. \quad (4.65)$$

From this, by l'Hôpital's rule, we have

$$K(x, x) = F_1'(x)G_1(x) + F_2'(x)G_2(x). \quad (4.66)$$

For $x > 0$ and $x \neq p^+$, we have

$$\begin{aligned} \frac{1}{A^+(x)}V\left(\frac{F_1}{x - p^+}\right)(x) &= \int_{\mathbb{R}_-} \frac{A^-(y)F_1(y)}{(x - y)(y - p^+)} dy \\ &= \frac{1}{x - p^+} \int_{\mathbb{R}_-} \left(\frac{A^-(y)F_1(y)}{x - y} - \frac{A^-(y)F_1(y)}{p^+ - y} \right) dy \\ &= \frac{1}{x - p^+} \left[\frac{(VF_1)(x)}{A^+(x)} - \frac{(VF_1)(p^+)}{A^+(p^+)} \right] \\ &= \frac{1}{x - p^+} \left(\frac{A^+(x) - F_1(x)}{A^+(x)} - \frac{A^+(p^+) - F_1(p^+)}{A^+(p^+)} \right) \\ &= \frac{1}{x - p^+} \left(\frac{-F_1(x)}{A^+(x)} + \frac{F_1(p^+)}{A^+(p^+)} \right). \end{aligned}$$

Similarly, if $x > 0$ and $x \neq p^+$, then

$$\begin{aligned} \frac{1}{A^+(x)}V\left(\frac{F_2}{x - p^+}\right)(x) &= \frac{1}{x - p^+} \left[\frac{(VF_2)(x)}{A^+(x)} - \frac{(VF_2)(p^+)}{A^+(p^+)} \right] \\ &= \frac{1}{x - p^+} \left(\frac{-F_2(x)}{A^+(x)} + \frac{F_2(p^+)}{A^+(p^+)} \right). \end{aligned}$$

We thus get

$$V\left(\frac{x-p^-}{x-p^+}K(x,p^-)\right)(x) = -\frac{x-p^-}{x-p^+}K(x,p^-) + \frac{p^+-p^-}{x-p^+}\frac{K(p^+,p^-)}{A^+(p^+)}A^+(x). \quad (4.67)$$

Now note that we have

$$\frac{y-p^-}{(x-y)(y-p^+)^2} = \frac{x-p^-}{(x-p^+)^2}\frac{1}{x-y} - \frac{x-p^-}{(x-p^+)^2}\frac{1}{p^+-y} + \frac{p^+-p^-}{x-p^+}\frac{1}{(p^+-y)^2}. \quad (4.68)$$

Given $f \in L^2(\mathbb{R}_-, dx)$, we have

$$\int_{\mathbb{R}_-} \frac{A^-(y)f(y)}{(x-y)^2} dy = -\frac{d}{dx} \int_{\mathbb{R}_-} \frac{A^-(y)f(y)}{x-y} dy = -\frac{d}{dx} \left[\frac{V(f)}{A^+} \right] (x). \quad (4.69)$$

Applying identities (4.68) and (4.69) and by denoting

$$H(x) = F_1(x)G_1(p^+) + F_2(x)G_2(p^+),$$

we get

$$\begin{aligned} \frac{V\left(\frac{x-p^-}{x-p^+}K(x,p^+)\right)(x)}{A^+(x)} &= \int_{\mathbb{R}_-} \frac{(y-p^-)A^-(y)H(y)}{(x-y)(y-p^+)^2} dy \\ &= \frac{x-p^-}{(x-p^+)^2} \int_{\mathbb{R}_-} \frac{A^-(y)H(y)}{x-y} dy - \frac{x-p^-}{(x-p^+)^2} \int_{\mathbb{R}_-} \frac{A^-(y)H(y)}{p^+-y} dy \\ &\quad + \frac{p^+-p^-}{x-p^+} \int_{\mathbb{R}_-} \frac{A^-(y)H(y)}{(p^+-y)^2} dy \\ &= \frac{x-p^-}{(x-p^+)^2} \left[\frac{V(H)(x)}{A^+(x)} - \frac{V(H)(p^+)}{A^+(p^+)} \right] - \frac{p^+-p^-}{x-p^+} \frac{d}{dx} \left[\frac{V(H)}{A^+} \right] (p^+). \end{aligned}$$

By (4.64), for $x > 0$, we have

$$\begin{cases} \frac{d}{dx} \left(\frac{V(F_1)}{A^+} \right) (x) = \frac{F'_1(x)A^+(x) - F_1(x)\frac{d}{dx}A^+(x)}{A^+(x)^2} \\ \frac{d}{dx} \left(\frac{V(F_2)}{A^+} \right) (x) = \frac{F'_2(x)A^+(x) - F_2(x)\frac{d}{dx}A^+(x)}{A^+(x)^2} \end{cases}.$$

Keeping in mind that the identities (4.65) and (4.66) hold, we obtain

$$V\left(\frac{x-p^-}{x-p^+}K(x,p^+)\right)(x) = -\frac{x-p^-}{x-p^+}K(x,p^+) + \frac{p^+-p^-}{x-p^+}\frac{K(p^+,p^+)}{A^+(p^+)}A^+(x). \quad (4.70)$$

Combining identities (4.67) and (4.70), we get

$$\begin{aligned} &V\left(\frac{x-p^-}{x-p^+}\left[K(p^+,p^+)K(x,p^-) - K(x,p^+)K(p^+,p^-)\right]\right)(x) \\ &= -\frac{x-p^-}{x-p^+}\left\{K(p^+,p^+)K(x,p^-) - K(x,p^+)K(p^+,p^-)\right\}. \end{aligned} \quad (4.71)$$

Since

$$K(p^+, p^+)K^{p^+}(x, p^-) = K(p^+, p^+)K(x, p^-) - K(p^+, p^-)K(x, p^+),$$

the identities (4.63) and (4.71) are equivalent. Thus we complete the proof of Step 2.

Remark 4.3. Denote

$$\psi(x) := \frac{x - p^-}{x - p^+} \varphi(x) = \frac{x - p^-}{x - p^+} \operatorname{sgn}(\cdot) K_L^{p^+}(\cdot, p^-),$$

then we can show that the identity (4.62) is equivalent to $\psi_+ = K_L(\psi)$, which is in turn equivalent to $\psi = \widehat{K}_L(\psi)$. Hence in Step 2, we in fact proved that

$$\frac{x - p^-}{x - p^+} \operatorname{sgn}(\cdot) K_L^{p^+}(\cdot, p^-) \in \operatorname{Ran}(\widehat{K}_L).$$

Step 3: Now we want to prove that $\operatorname{Ran}(\widehat{K}_L^p) \supset \operatorname{Ran}(\widehat{K}_{g_p L g_p})$, this is equivalent to

$$\operatorname{Ran}(\widehat{K}_L^p)^\perp \subset \operatorname{Ran}(\widehat{K}_{g_p L g_p})^\perp. \quad (4.72)$$

By using (4.54), we have

$$\operatorname{Ran}(\widehat{K}_L^p)^\perp = \left(\operatorname{Ran}(\widehat{K}_L)^\perp \ominus \mathbb{C} \operatorname{sgn}(\cdot) K_L(\cdot, p^-) \right) \oplus \mathbb{C} \operatorname{sgn}(\cdot) K_L^{p^-}(\cdot, p^+).$$

From (4.57), we know that

$$\begin{aligned} \operatorname{Ran}(\widehat{K}_L)^\perp &= \left\{ f_1 \oplus (-V^* f_1) : f_1 \in L^2(\mathbb{R}_+, dx) \right\}, \\ \operatorname{Ran}(\widehat{K}_{g_p L g_p})^\perp &= \left\{ f_2 \oplus \left(-g_p^- V^*(g_p^+ f_2) \right) : f_2 \in L^2(\mathbb{R}_+, dx) \right\}. \end{aligned}$$

Now after switching the rôles of $L^2(\mathbb{R}_-, dx)$ and $L^2(\mathbb{R}_+, dx)$; the kernels V and $-V^*$; the pairs (p^+, p^-) and (p^-, p^+) ; the functions $\frac{x-p^+}{x-p^-}$ and $\frac{x-p^-}{x-p^+}$; and also the pairs of vectors $(\operatorname{sgn}(\cdot) K_L(\cdot, p^+), \operatorname{sgn}(\cdot) K_L^{p^+}(\cdot, p^-))$ and $(\operatorname{sgn}(\cdot) K_L(\cdot, p^-), \operatorname{sgn}(\cdot) K_L^{p^-}(\cdot, p^+))$, we arrive exactly at the same situation as above in proving $\operatorname{Ran}(\widehat{K}_L^p) \subset \operatorname{Ran}(\widehat{K}_{g_p L g_p})$. Hence we may obtain (4.72) by repeating the same arguments in Step 1 and Step 2. \square

4.2 Sufficient condition for equivalence of two L -processes

In this section, we formulate a sufficient condition for two L -processes to be equivalent on the level of their L -kernels.

Lemma 4.6. *Let L be an operator satisfying Condition 4. Then the operator K_L is a J -self-adjoint operator satisfying all the conditions of Theorem 1.3. In particular, K_L is the correlation kernel of a determinantal point process on \mathbb{R}^* .*

Lemma 4.7. *Let L be an operator satisfying Condition 4. Let $g : \mathbb{R} \rightarrow \mathbb{R}$ be a bounded Borel function. Then the operator gLg satisfies also Condition 4. Moreover, the operator $K_{gLg} = gLg(1 + gLg)^{-1}$ is given by*

$$K_{gLg} = gK_L(1 + (g^2 - 1)K_L)^{-1}g. \quad (4.73)$$

Condition 5. Assume that L is a kernel given as in (4.55) such that the operator $V : L^2(\mathbb{R}_-) \rightarrow L^2(\mathbb{R}_+)$ satisfying the following condition: for any $\varepsilon > 0$, the operators $\chi_{(\varepsilon, \infty)}V$, $V\chi_{(-\infty, -\varepsilon)}$, $\chi_{(0, \varepsilon)}xV$ and $Vy\chi_{(-\varepsilon, 0)}$ are Hilbert-Schmidt.

Lemma 4.8. *An L -kernel satisfying Condition 3 is an L -kernel satisfying Condition 5.*

Recall that we denote by μ_L the determinantal point process on \mathbb{R}^* whose correlation kernel is $K_L = L(1 + L)^{-1}$.

Theorem 4.9. *Let L be a kernel satisfying Condition 5. Assume that g is a bounded real function such that $|g(x) - 1| \leq C|x|$ and there exists $\varepsilon \in (0, 1)$ such that the subset $\{x \in \mathbb{R}^* : |g(x) - 1| > \varepsilon\}$ has a positive distance from 0. Then the following limit*

$$\overline{S}_L[\log g^2](X) := \lim_{\delta \rightarrow 0^+} \sum_{x \in X, |x| \geq \delta} \log g(x)^2 - \mathbb{E}_{\mu_L} \left(\sum_{x \in X, |x| \geq \delta} \log g(x)^2 \right)$$

exists for μ_L -almost every configuration $X \in \text{Conf}(\mathbb{R}^)$. Moreover, we have*

$$\exp(\overline{S}_L[\log g^2]) \in L^1(\text{Conf}(\mathbb{R}^*), \mu_L)$$

and we have

$$\mu_{gLg}(dX) = \frac{\exp(\overline{S}_L[\log g^2](X))}{\mathbb{E}_{\mu_L}[\exp(\overline{S}_L[\log g^2])]} \cdot \mu_L(dX).$$

4.3 Remarks on Conditions 3, 4, 5 and Proofs of Lemmas 4.6, 4.7, 4.8

Remark 4.4. Let L be an operator as in Condition 4, then K_L has the following block form:

$$K_L = \begin{bmatrix} VV^*(1 + VV^*)^{-1} & V(1 + V^*V)^{-1} \\ -V^*(1 + VV^*)^{-1} & V^*V(1 + V^*V)^{-1} \end{bmatrix}. \quad (4.74)$$

Remark 4.5. By the block form (4.74) of the operator K_L and the following elementary order inequalities for positive operators

$$\frac{1}{1 + \|V\|^2} VV^* \leq VV^*(1 + VV^*)^{-1} \leq VV^*$$

and

$$\frac{1}{1 + \|V\|^2} V^* V \leq V^* V (1 + V^* V)^{-1} \leq V^* V,$$

we may formulate Condition 5 in terms of the kernel K_L as follows: for any $\varepsilon > 0$, we have

$$\int_{|x| \geq \varepsilon} K_L(x, x) dx < \infty \text{ and } \int_{|x| \leq \varepsilon} x^2 K_L(x, x) dx < \infty. \quad (4.75)$$

In particular, by [Sos00, Thm. 4], the first inequality in (4.75) implies that the relation

$$\#\mathbb{R} \setminus (-\varepsilon, \varepsilon)(X) < \infty \quad (4.76)$$

holds for \mathbb{P}_{K_L} -almost every configuration $X \in \text{Conf}(\mathbb{R}^*)$.

Remark 4.6. Finite rank perturbation of the kernel K_L will not affect the inequalities (4.75).

Remark 4.7. In Condition 3, we require the boundedness of L and hence of V . In general, an operator V given by a kernel

$$V(x, y) = \frac{A^+(x)A^-(y)}{x - y}$$

such that $A \in L^2(\mathbb{R}^*, dx)$, is not necessarily bounded. See e.g., Propositions 2.2 and 2.3 in [Ols].

Remark 4.8. The operator norm of the operator K_L given in (4.74) satisfies

$$\|K_L\| < 1. \quad (4.77)$$

Indeed, by Proposition 4.5, the operator K_L is J -self-adjoint and \widehat{K} is an orthogonal projection, by [Lyt13, Prop. 7], we have $\|K_L\| \leq 1$. We shall exclude the possibility $\|K_L\| = 1$. Indeed, if this were the case (i.e., $\|K_L\| = 1$), then by [Lyt13, Prop. 8], we would get

$$\left\| \begin{bmatrix} VV^*(1 + VV^*)^{-1} & 0 \\ 0 & V^*V(1 + V^*V)^{-1} \end{bmatrix} \right\| = 1. \quad (4.78)$$

However, by functional calculus, we have

$$\|VV^*(1 + VV^*)^{-1}\| = \|V^*V(1 + V^*V)^{-1}\| = \frac{\|V\|^2}{1 + \|V\|^2} < 1.$$

Hence (4.78) is not valid. This completes the proof of (4.77).

We will often use

Remark 4.9. Let a, b be two bounded linear operator on a Hilbert space. If $1 + ab$ is invertible, then so is $1 + ba$. We have

$$a(1 + ba)^{-1} = (1 + ab)^{-1}a. \quad (4.79)$$

Proof of Lemma 4.8. Let L be a kernel satisfying Condition 3. Let $\varepsilon > 0$. The simple estimate:

$$\begin{aligned} & \max \left(\int_{\varepsilon}^{\infty} dx \int_{-\infty}^0 \left| \frac{A^+(x)A^-(y)}{x-y} \right|^2 dy, \int_0^{\infty} dx \int_{-\infty}^{-\varepsilon} \left| \frac{A^+(x)A^-(y)}{x-y} \right|^2 dy \right) \\ & \leq \frac{\|A^+\|_2^2 \|A^-\|_2^2}{\varepsilon^2} \end{aligned}$$

shows that the operators with the kernels

$$\chi_{(\varepsilon, \infty)}(x) \frac{A^+(x)A^-(y)}{x-y} \text{ and } \frac{A^+(x)A^-(y)}{x-y} \chi_{(-\infty, -\varepsilon)}(y)$$

are Hilbert-Schmidt.

The inequality:

$$\begin{aligned} & \max \left(\int_0^{\varepsilon} dx \int_{-\infty}^0 x^2 \left| \frac{A^+(x)A^-(y)}{x-y} \right|^2 dy, \int_0^{\infty} dx \int_{-\varepsilon}^0 y^2 \left| \frac{A^+(x)A^-(y)}{x-y} \right|^2 dy \right) \\ & \leq \|A^+\|_2^2 \|A^-\|_2^2 \end{aligned}$$

implies that $\chi_{(0, \varepsilon)}xV$ and $Vy\chi_{(-\varepsilon, 0)}$ are also Hilbert-Schmidt.

The Lemma 4.8 is proved completely. \square

Proof of Lemma 4.6. By Proposition 4.5, the operator \widehat{K}_L is self-adjoint, hence K_L is J -self-adjoint. By Remark 4.4, the operators $P_+K_LP_+$ and $P_-K_LP_-$ are non-negative.

Let Δ_1, Δ_2 be compact subsets of \mathbb{R}^* such that $\Delta_1 \subset \mathbb{R}_+$ and $\Delta_2 \subset \mathbb{R}_-$. We now check that the operators $\chi_{\Delta_i}K_L\chi_{\Delta_i}$ ($i = 1, 2$) are in trace-class. Let us verify this for $i = 1$. Since VV^* is positive, we have

$$VV^*(1 + VV^*)^{-1} \leq VV^*,$$

it follows that

$$0 \leq \chi_{\Delta_1}K_L\chi_{\Delta_1} = \chi_{\Delta_1}VV^*(1 + VV^*)^{-1}\chi_{\Delta_1} \leq \chi_{\Delta_1}VV^*\chi_{\Delta_1}.$$

The assumption that $\chi_{\Delta_1}V$ is a Hilbert-Schmidt operator now implies that $\chi_{\Delta_1}K_L\chi_{\Delta_1}$ is a trace-class operator. The case of $i = 2$ is similar.

Finally, let us verify that $\chi_{\Delta_2}K_L\chi_{\Delta_1}$ is a Hilbert-Schmidt operator. Indeed, we have

$$\chi_{\Delta_2}K_L\chi_{\Delta_1} = -\chi_{\Delta_2}V^*(1 + VV^*)^{-1}\chi_{\Delta_1} = -\chi_{\Delta_2}(1 + V^*V)^{-1}V^*\chi_{\Delta_1}.$$

Since the space of Hilbert-Schmidt operators is an ideal of the algebra $\mathcal{L}(L^2(\mathbb{R}))$ of all bounded linear operators, the assumption that $V^*\chi_{\Delta_1} = (\chi_{\Delta_1}V)^*$ is a Hilbert-Schmidt operator implies that $\chi_{\Delta_2}K_L\chi_{\Delta_1}$ is a Hilbert-Schmidt operator. The proof is complete. \square

Proof of Lemma 4.7. We have

$$gLg = \begin{bmatrix} 0 & g\chi_{\mathbb{R}_+}Vg\chi_{\mathbb{R}_-} \\ -g\chi_{\mathbb{R}_-}V^*g\chi_{\mathbb{R}_+} & 0 \end{bmatrix}.$$

Since g is bounded, the operator $g\chi_{\mathbb{R}_+}Vg\chi_{\mathbb{R}_-}$ satisfies all the conditions in Condition 4 imposed on the operator V .

Set $a = g$ and $b = gK_L(1 - K_L)^{-1}$. The

$$1 + gLg = 1 + gK_L(1 - K_L)^{-1}g = 1 + ba$$

is invertible. By Remark 4.9, the operator $1 + ab = 1 + g^2K_L(1 - K_L)^{-1}$ is also invertible. The identity (4.79) now yields

$$\begin{aligned} K_{gLg} &= gK_L(1 - K_L)^{-1}g(1 + gK_L(1 - K_L)^{-1}g)^{-1} \\ &= gK_L(1 - K_L)^{-1}(1 + g^2K_L(1 - K_L)^{-1})^{-1}g \\ &= gK_L(1 - K_L + g^2K_L)^{-1}g, \end{aligned}$$

which is the desired identity. The proof is complete. \square

4.4 A preliminary version of Theorem 4.9

As usual, given a function h defined on \mathbb{R}^* , we define the multiplicative functional $\Psi[h] : \text{Conf}(\mathbb{R}^*) \rightarrow \mathbb{R}$ by the following formula

$$\Psi[h](X) = \prod_{x \in X} h(x), \quad X \in \text{Conf}(\mathbb{R}^*), \quad (4.80)$$

provided the right-hand side converges absolutely.

Proposition 4.10. *Let L be an operator satisfying Condition 5. If g is a bounded real function defined on \mathbb{R}^* such that $\text{supp}(g^2 - 1)$ has a positive distance from the origin. Then*

$$\mu_{gLg} = \frac{\Psi[g^2]}{\mathbb{E}_{\mu_L}(\Psi[g^2])} \cdot \mu_L.$$

Proof. By general theory on determinantal measures, it suffices to prove that for any continuous real function h such that $\text{supp}(h - 1)$ has a positive distance from the origin, we have

$$\mathbb{E}_{\mu_{gLg}}(\Psi[h]) = \frac{\mathbb{E}_{\mu_L}(\Psi[h]\Psi[g^2])}{\mathbb{E}_{\mu_L}(\Psi[g^2])} = \frac{\mathbb{E}_{\mu_L}(\Psi[hg^2])}{\mathbb{E}_{\mu_L}(\Psi[g^2])}. \quad (4.81)$$

Fix such a function h , set $\Delta = \text{supp}(h - 1) \cup \text{supp}(g^2 - 1)$. Then Δ also has a positive distance from the origin. Obviously, $\text{supp}(hg^2 - 1) \subset \Delta$. By definition of determinantal point process and Theorem 2.7, we have

$$\begin{aligned}\mathbb{E}_{\mu_{gLg}}(\Psi[h]) &= \det(1 + (h - 1)\chi_\Delta \cdot K_{gLg} \cdot \chi_\Delta); \\ \mathbb{E}_{\mu_L}(\Psi[hg^2]) &= \det(1 + (hg^2 - 1)\chi_\Delta \cdot K_L \cdot \chi_\Delta); \\ \mathbb{E}_{\mu_L}(\Psi[g^2]) &= \det(1 + (g^2 - 1)\chi_\Delta \cdot K_L \cdot \chi_\Delta).\end{aligned}$$

Recall that by Lemma 4.7, the operator $K(g)$ satisfies all the conditions of Theorem 1.3 and is given by $K_{gLg} = gK_L(1 + (g^2 - 1)K_L)^{-1}g$. By (2.22), we have

$$K_L^\Delta = \chi_\Delta \cdot K_L \cdot \chi_\Delta \in \mathcal{L}_{1|2}(L^2(\mathbb{R})) \subset \mathcal{L}_2(L^2(\mathbb{R})).$$

By Remark (4.9), we get

$$\begin{aligned}\chi_\Delta \cdot K_{gLg} \cdot \chi_\Delta &= g\chi_\Delta K_L(1 + (g^2 - 1)K_L)^{-1}\chi_\Delta g \\ &= g\chi_\Delta K_L \left[(1 + \chi_\Delta(g^2 - 1)\chi_\Delta K_L)^{-1}\chi_\Delta \right] g \\ &= g\chi_\Delta K_L \left[\chi_\Delta(1 + (g^2 - 1)\chi_\Delta K_L \chi_\Delta)^{-1} \right] g \\ &= gK_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1}g.\end{aligned}\tag{4.82}$$

Observe that we can write

$$\begin{aligned}&K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1} \\ &= \left[K_L^\Delta(1 + (g^2 - 1)K_L^\Delta) + K_L^\Delta(1 - g^2)K_L^\Delta \right] (1 + (g^2 - 1)K_L^\Delta)^{-1} \\ &= K_L^\Delta + K_L^\Delta(1 - g^2)K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1},\end{aligned}$$

then by Hölder inequality, we have

$$\|K_L^\Delta(1 - g^2)K_L^\Delta\|_1 \leq \|K_L^\Delta\|_2 \|(1 - g^2)K_L^\Delta\|_2 \leq \|g^2 - 1\|_\infty \|K_L^\Delta\|_2^2 < \infty,$$

that is, the operator $K_L^\Delta(1 - g^2)K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1}$ is in trace-class. It follows that

$$K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1} \in \mathcal{L}_{1|2}(L^2(\mathbb{R})).$$

Thus we have

$$\begin{aligned}\det(1 + (h - 1)\chi_\Delta K_{gLg} \chi_\Delta) &= \det(1 + (h - 1)gK_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1}g) \\ &= \det(1 + (h - 1)g^2K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1}).\end{aligned}$$

An application of the identity

$$1 + (h - 1)g^2K_L^\Delta(1 + (g^2 - 1)K_L^\Delta)^{-1} = \left[1 + (hg^2 - 1)K_L^\Delta \right] (1 + (g^2 - 1)K_L^\Delta)^{-1}$$

yields that

$$\det(1 + (h - 1)\chi_\Delta K_{gLg}\chi_\Delta) = \frac{\det(1 + (hg^2 - 1)K_L^\Delta)}{\det(1 + (g^2 - 1)K_L^\Delta)}.$$

This shows the desired identity (4.81). The proof of Proposition 4.10 is complete. \square

The following lemma will be useful for us.

Lemma 4.11. *Assume that L is an operator satisfying Condition 4.*

- *Let $\alpha : \mathbb{R}^* \rightarrow \mathbb{C}$ be any measurable function with non-zero constant modulus. Then $\alpha L \alpha^{-1}$ induces a determinantal measure $\mu_{\alpha L \alpha^{-1}} = \mathbb{P}_{K_{\alpha L \alpha^{-1}}}$ and $\mu_{\alpha L \alpha^{-1}} = \mu_L$.*
- *Let $\lambda \neq 0$ be a numerical constant and let c_λ be the function defined by*

$$c_\lambda = \lambda \chi_{\mathbb{R}_+} + \lambda^{-1} \chi_{\mathbb{R}_-}. \quad (4.83)$$

Then $c_\lambda L c_\lambda = L$.

Proof. By assumption $\alpha = a\gamma$ with $a = |\alpha| > 0$ a numerical constant and γ a measurable function with values in the unit circle. We shall see that $\alpha L \alpha^{-1}$ satisfies also Condition 4. Indeed,

$$\alpha L \alpha^{-1} = \gamma L \bar{\gamma} = \begin{bmatrix} 0 & \gamma^+ V \bar{\gamma}^- \\ -\gamma^- V^* \bar{\gamma}^+ & 0 \end{bmatrix} = \begin{bmatrix} 0 & \gamma^+ V \bar{\gamma}^- \\ -(\gamma^+ V \bar{\gamma}^-)^* & 0 \end{bmatrix}.$$

Hence the determinantal measure $\mu_{\alpha L \alpha^{-1}}$ is well-defined. The coincidence of $\mu_{\alpha L \alpha^{-1}}$ and μ_L is an easy consequence of the fact that $K_{\alpha L \alpha^{-1}} = \alpha K_L \alpha^{-1}$ and

$$\det(\alpha(x_i) K_L(x_i, x_j) \alpha(x_j)^{-1})_{1 \leq i, j \leq n} = \det(K_L(x_i, x_j))_{1 \leq i, j \leq n}.$$

The second assertion is an easy consequence of the following identity

$$\begin{bmatrix} \lambda & 0 \\ 0 & \lambda^{-1} \end{bmatrix} \begin{bmatrix} 0 & V \\ -V^* & 0 \end{bmatrix} \begin{bmatrix} \lambda & 0 \\ 0 & \lambda^{-1} \end{bmatrix} = \begin{bmatrix} 0 & V \\ -V^* & 0 \end{bmatrix}.$$

\square

4.5 Regularization of additive and multiplicative functionals

4.5.1 Additive functionals

Assume now that L is a kernel satisfying Condition 5. Recall that we set

$$\mu_L = \mathbb{P}_{K_L}.$$

Let $f : \mathbb{R}^* \rightarrow \mathbb{C}$ be a Borel function. Then we write

$$T[f](X) = S[f^\circ](X) = \sum_{x \in X} \text{sgn}(x) f(x), \quad (4.84)$$

provided the right hand side converges absolutely, otherwise, $T[f]$ is not defined at the configuration X .

If $T[f]$ is μ_L -almost surely defined and $T[f] \in L_1(\text{Conf}(\mathbb{R}^*), \mu_L)$, then we set

$$\overline{T}_L[f] := T[f] - \mathbb{E}_{\mu_L}(T[f]). \quad (4.85)$$

Following the idea in [Buf14], we will now provide a sufficient condition such that $\overline{T}_L[f]$ can be defined even when $T[f]$ is not. Set

$$\mathcal{V}_L(f) = \frac{1}{2} \iint_{\mathbb{R}^2} |f(x) - f(y)|^2 |K_L(x, y)|^2 dx dy. \quad (4.86)$$

Note that for any $\lambda \in \mathbb{C}$, we have $\mathcal{V}_L(f + \lambda) = \mathcal{V}_L(f)$. Note also that

$$\mathcal{V}_L(f) \leq \int_{\mathbb{R}} |f(x)|^2 K_L(x, x) dx. \quad (4.87)$$

By Lemma 3.2, if $T[f] \in L_2(\text{Conf}(\mathbb{R}^*), \mu_L)$, then $\mathcal{V}_L(f) < \infty$ and

$$\text{Var}_{\mu_L}(T[f]) = \mathbb{E}_{\mu_L} |\overline{T}_L[f]|^2 = \mathcal{V}_L(f). \quad (4.88)$$

Definition 4.3. Let $\mathcal{N}_0(L)$ be the linear space of Borel functions $f : \mathbb{R}^* \rightarrow \mathbb{C}$ such that there exist $\varepsilon > 0$, depending on f , so that

$$\text{supp}(f) \subset \{x \in \mathbb{R} : |x| \geq \varepsilon\} \text{ and } \int_{\mathbb{R}} |f(x)|^2 K_L(x, x) dx < \infty.$$

Definition 4.4. Let $\mathcal{N}(L)$ be the linear space of Borel functions $f : \mathbb{R}^* \rightarrow \mathbb{C}$ such that

$$\mathcal{V}_L(f) = \frac{1}{2} \iint_{\mathbb{R}^2} |f(x) - f(y)|^2 |K_L(x, y)|^2 dx dy < \infty; \quad (4.89)$$

$$\int_{|x| \geq \varepsilon} |f(x)|^2 K_L(x, x) dx < \infty, \text{ for all } \varepsilon > 0; \quad (4.90)$$

$$\lim_{\varepsilon \rightarrow 0^+} \iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |f(x)|^2 |K_L(x, y)|^2 dx dy = 0. \quad (4.91)$$

We endow the linear space $\mathcal{N}(L)$ with a Hilbert space structure $d_{\mathcal{N}(L)}$ by the formula

$$d_{\mathcal{N}(L)}(f, g) = \|f - g\|_{\mathcal{N}_L} := \sqrt{\mathcal{V}_L(f - g)}.$$

Remark 4.10. If $f \in \mathcal{N}_0(L)$, then by Cauchy-Buniakovsky-Schwarz inequality and the first inequality in (4.75), we have

$$\int_{\mathbb{R}} |f(x)| K_L(x, x) dx \leq \left(\int_{\mathbb{R}} |f(x)|^2 K_L(x, x) dx \cdot \int_{\text{supp}(f)} K_L(x, x) dx \right)^{1/2} < \infty.$$

This means that $T[f] \in L_1(\text{Conf}(\mathbb{R}^*), \mu_L)$, hence $\overline{T}_L[f]$ is well-defined by formula (4.85). Moreover, by the relations (4.87) and (4.88), we actually have

$$\overline{T}_L[f] \in L_2(\text{Conf}(\mathbb{R}^*), \mu_L).$$

Proposition 4.12. *We have the inclusion*

$$\mathcal{N}_0(L) \subset \mathcal{N}(L).$$

Moreover, $\mathcal{N}_0(L)$ is dense in $\mathcal{N}(L)$. More precisely, if f is a function in $\mathcal{N}(L)$, then for any $\varepsilon > 0$, the truncated function $f \chi_{\mathbb{R} \setminus (-\varepsilon, \varepsilon)}$ is in $\mathcal{N}_0(L)$ and we have

$$\lim_{\varepsilon \rightarrow 0^+} \mathcal{V}_L(f \chi_{\mathbb{R} \setminus (-\varepsilon, \varepsilon)} - f) = 0. \quad (4.92)$$

Proof. The inclusion $\mathcal{N}_0(L) \subset \mathcal{N}(L)$ follows from their definitions and the following inequality

$$\iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |f(x)|^2 |K_L(x, y)|^2 dx dy \leq \int_{|x| \leq \varepsilon} |f(x)|^2 K_L(x, x) dx,$$

By definition of $\mathcal{N}(L)$, we have $f \chi_{\mathbb{R} \setminus (-\varepsilon, \varepsilon)} \in \mathcal{N}_0(L)$. Since

$$\begin{aligned} \mathcal{V}_L(f \chi_{\mathbb{R} \setminus (-\varepsilon, \varepsilon)} - f) &= \frac{1}{2} \iint_{\mathbb{R}^2} |f \chi_{[-\varepsilon, \varepsilon]}(x) - f \chi_{[-\varepsilon, \varepsilon]}(y)|^2 |K_L(x, y)|^2 dx dy \\ &\leq \iint_{|x| \leq \varepsilon, |y| \leq \varepsilon} |f(x) - f(y)|^2 |K_L(x, y)|^2 dx dy + \iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |f(x)|^2 |K_L(x, y)|^2 dx dy \\ &\quad + \iint_{|x| \geq \varepsilon, |y| \leq \varepsilon} |f(y)|^2 |K_L(x, y)|^2 dx dy \\ &= \iint_{|x| \leq \varepsilon, |y| \leq \varepsilon} |f(x) - f(y)|^2 |K_L(x, y)|^2 dx dy + 2 \iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |f(x)|^2 |K_L(x, y)|^2 dx dy. \end{aligned}$$

By the assumption $\mathcal{V}_L(f) < \infty$ and the relation (4.91), we get the desired relation (4.92). \square

Proposition 4.13. *The isometric embedding*

$$\begin{array}{ccc} \overline{T} : \mathcal{N}_0(L) & \hookrightarrow & L_2(\text{Conf}(\mathbb{R}^*), \mu_L) \\ f & \mapsto & \overline{T}_L[f] \end{array}$$

extends uniquely to an isometric embedding $\mathcal{N}(L) \hookrightarrow L_2(\text{Conf}(\mathbb{R}^), \mu_L)$.*

Definition 4.5. Given a function $f \in \mathcal{N}(L)$, by slightly abusing the notation, we denote by $\overline{T}_L[f]$ the image of f under the embedding map $\mathcal{N}(L) \hookrightarrow L_2(\text{Conf}(\mathbb{R}^*), \mu_L)$ in Proposition 4.13. We will call $\overline{T}_L[f]$ the normalized twisted additive functional corresponding to f and μ_L .

Remark 4.11. For all $f \in \mathcal{N}(L)$, we have $\mathbb{E}_{\mu_L}(\overline{T}_L[f]) = 0$.

Remark 4.12. If $f \in \mathcal{N}(L)$, then up to passing to a sequence ε_n tending to zero if necessary, we may write the following *pointwise* relation: for μ_L -almost every configuration $X \in \text{Conf}(\mathbb{R}^*)$,

$$\overline{T}_L[f](X) = \lim_{\varepsilon \rightarrow 0^+} \left(\sum_{x \in X, |x| \geq \varepsilon} \text{sgn}(x)f(x) - \mathbb{E}_{\mu_L} \sum_{x \in X, |x| \geq \varepsilon} \text{sgn}(x)f(x) \right). \quad (4.93)$$

4.5.2 Multiplicative functionals

Notation. Let $f : \mathbb{R}^* \rightarrow \mathbb{C}$ be a measurable function, denote

$$f^+ = f\chi_{\mathbb{R}_+} \text{ and } f^- = f\chi_{\mathbb{R}_-}.$$

If the essential support $\text{supp}(f^-)$ of the function f^- is the whole negative semi-axis \mathbb{R}_- , then we may define

$$f^\vee(x) := f^+(x) + (f^-(x))^{-1}. \quad (4.94)$$

Definition 4.6. Given a function $g : \mathbb{R}^* \rightarrow [0, \infty]$ such that $\{x \in \mathbb{R}^* : g(x) = 0\}$ is Lebesgue negligible and $\log g \in \mathcal{N}(L)$, then we set

$$\widetilde{\Psi}_L[g] = \exp(\overline{T}_L[\log(g^\vee)]),$$

where by definition (4.94), $g^\vee(x) := g(x)\chi_{\mathbb{R}_+}(x) + g(x)^{-1}\chi_{\mathbb{R}_-}(x)$. If moreover, $\mathbb{E}_{\mu_L}\widetilde{\Psi}_L[g]$ is finite, then we define

$$\overline{\Psi}_L[g] = \frac{\widetilde{\Psi}_L[g]}{\mathbb{E}_{\mu_L}\widetilde{\Psi}_L[g]}.$$

Remark 4.13. If g is a function such that $\log g \in \mathcal{N}(L)$, then $\mathbb{E}_{\mu_L}\widetilde{\Psi}_L[g] \in [1, \infty]$. Indeed, by Jensen's inequality and Remark 4.11, we have

$$\mathbb{E}_{\mu_L}\widetilde{\Psi}_L[g] = \mathbb{E}_{\mu_L} \exp(\overline{T}_L[\log(g^\vee)]) \geq \exp(\mathbb{E}_{\mu_L}(\overline{T}_L \log g^\vee)) = 1. \quad (4.95)$$

The formalism of regularized multiplicative functional $\overline{\Psi}_L[g]$ now allows us state the following

Theorem 4.14. *Let $g : \mathbb{R}^* \rightarrow [0, \infty)$ be a non-negative bounded function. Assume that there exists $\varepsilon \in (0, 1)$ such that $E_\varepsilon = \{x \in \mathbb{R}^* : |g(x)^2 - 1| > \varepsilon\}$ has a positive distance from the origin and*

$$\int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx < \infty. \quad (4.96)$$

Then $\log g \in \mathcal{N}(L)$ and $\tilde{\Psi}_L[g^2] \in L_1(\text{Conf}(\mathbb{R}^), \mu_L)$. Moreover, we have*

$$\mu_{gLg} = \overline{\Psi}_L[g^2] \cdot \mu_L. \quad (4.97)$$

4.6 Proof of Theorem 4.14

Definition 4.7. Let $\mathcal{M}_2(L)$ denote the set of functions g on \mathbb{R} such that

- (1) $0 < \inf_{\mathbb{R}} g \leq \sup_{\mathbb{R}} g < \infty$;
- (2) $\int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx < \infty$.

Recall that by definition (4.94), to a function g , we assign g^\vee in the following way:

$$g^\vee(x) := g(x)\chi_{\mathbb{R}_+}(x) + g(x)^{-1}\chi_{\mathbb{R}_-}(x).$$

Proposition 4.15. *Let $g \in \mathcal{M}_2(L)$. Then $\log g$ and $\log(g^\vee)$ are functions in $\mathcal{N}(L)$. In particular, the functional $\tilde{\Psi}_L[g] = \exp(\overline{T}_L[\log(g^\vee)])$ is well-defined. Moreover, we have*

$$\mu_{gLg} = \overline{\Psi}_L[g^2] \cdot \mu_L. \quad (4.98)$$

We postpone its proof to the next section.

Now we are in a position to prove Theorem 4.14. But first, let us note that for a function g as in Proposition 4.10, the regularized multiplicative functional $\overline{\Psi}_L[g^2]$ defined as above is also expressed by $\overline{\Psi}_L[g^2] = C\Psi[g^2]$ for a certain constant $C > 0$.

Proof of Theorem 4.14. Let $\varepsilon \in (0, 1)$ be such that $\{x \in \mathbb{R}^* : |g(x) - 1| > \varepsilon\}$ has a positive distance from the origin. Set g_1, g_2 to be two positive functions determined by

$$g_1 = (g - 1)\chi_{\{x \in \mathbb{R}^* : |g(x) - 1| \leq \varepsilon\}} + 1. \quad (4.99)$$

$$g_2 = (g - 1)\chi_{\{x \in \mathbb{R}^* : |g(x) - 1| > \varepsilon\}} + 1. \quad (4.100)$$

By definition, $g = g_1 g_2$. Note that $1 - \varepsilon \leq \inf_{\mathbb{R}} g_1 \leq \sup_{\mathbb{R}} g_1 \leq 1 + \varepsilon$. This combining with assumption (4.96) shows that the function g_1 is in $\mathcal{M}_2(L)$. Hence by Proposition 4.15, we have

$$\mu_{g_1 L g_1} = \overline{\Psi}_L[g_1^2] \cdot \mu_L. \quad (4.101)$$

Now since $\text{supp}(g_2 - 1)$ has a positive distance from the origin and $gLg = g_2(g_1Lg_1)g_2$, by Proposition 4.10, we have

$$\mu_{gLg} = \frac{\Psi[g_2^2]}{\mathbb{E}_{\mu_{g_1Lg_1}} \Psi[g_2^2]} \cdot \mu_{g_1Lg_1}. \quad (4.102)$$

Combining (4.101) and (4.102), we get

$$\mu_{gLg} = \frac{\Psi[g_2^2]}{\mathbb{E}_{\mu_{g_1Lg_1}} \Psi[g_2^2]} \cdot \bar{\Psi}_L[g_1^2] \cdot \mu_L. \quad (4.103)$$

Since

$$\frac{\Psi[g_2^2]}{\mathbb{E}_{\mu_{g_1Lg_1}} \Psi[g_2^2]} \cdot \bar{\Psi}_L[g_1^2] = C_1 \tilde{\Psi}_L[g_2^2] \cdot C_2 \tilde{\Psi}_L[g_1^2] = C_1 C_2 \tilde{\Psi}_L[g_1^2 g_2^2] = C_1 C_2 \tilde{\Psi}_L[g^2],$$

and

$$\int C_1 C_2 \tilde{\Psi}_L[g^2] d\mu_L = 1,$$

we get

$$\frac{\Psi[g_2^2]}{\mathbb{E}_{\mu_{g_1Lg_1}} \Psi[g_2^2]} \cdot \bar{\Psi}_L[g_1^2] = \bar{\Psi}_L[g^2],$$

hence we complete the proof of the desired relation (4.97). \square

4.7 Proof of Proposition 4.15

Let us endow $\mathcal{M}_2(L)$ with a metric $d_{\mathcal{M}_2(L)}$ by setting

$$d_{\mathcal{M}_2(L)}(g_1, g_2) = \sqrt{\int_{\mathbb{R}} |g_1(x) - g_2(x)|^2 K_L(x, x) dx}.$$

By definition, $\mathcal{M}_2(L)$ is a semigroup under pointwise multiplication. Clearly, if g is a function in $\mathcal{M}_2(L)$, then so is g^\vee .

We shall first prove the following

Lemma 4.16. *Let $g \in \mathcal{M}_2(L)$. Then $\log g$ and $\log(g^\vee)$ are functions in $\mathcal{N}(L)$.*

Proof of Lemma 4.16. Assume that $g \in \mathcal{M}_2(L)$. Then there exist $c, C > 0$ such that $c \leq g(x) \leq C$. The boundedness of the function $\log g$ combining with the assumption (4.75) yields the inequality (4.90) for $\log g$.

Now since the function $|\log t - (t - 1)|/(t - 1)^2$ is bounded on the interval $[c, C]$, there exists $C' > 0$ such that

$$|\log g(x) - (g(x) - 1)| \leq C'(g(x) - 1)^2. \quad (4.104)$$

By taking $C'' = 1 + C' \max(|C - 1|, |c - 1|)$, we have

$$|\log g(x)| \leq C'' |g(x) - 1|. \quad (4.105)$$

It follows that

$$\int_{\mathbb{R}} |\log g(x)|^2 K_L(x, x) dx \leq (C'')^2 \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx < \infty.$$

Hence by applying (4.87), we have $\mathcal{V}_L(\log g) < \infty$. Following from (4.105), we also have

$$\begin{aligned} & \limsup_{\varepsilon \rightarrow 0^+} \iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |\log g(x)|^2 |K_L(x, y)|^2 dx dy \\ & \leq \lim_{\varepsilon \rightarrow 0^+} (C'')^2 \iint_{|x| \leq \varepsilon, |y| \geq \varepsilon} |g(x) - 1|^2 |K_L(x, y)|^2 dx dy \\ & \leq \lim_{\varepsilon \rightarrow 0^+} (C'')^2 \int_{|x| \leq \varepsilon} |g(x) - 1|^2 K_L(x, x) dx = 0. \end{aligned}$$

This completes the proof that $\log g \in \mathcal{N}(L)$. The same argument for $\log(g^\vee)$ since $g \in \mathcal{M}_2(L)$ implies that $g^\vee \in \mathcal{M}_2(L)$. \square

Proposition 4.17. *If $g \in \mathcal{M}_2(L)$, then $\tilde{\Psi}_L[g] \in L_1(\text{Conf}(\mathbb{R}^*), \mu_L)$. Moreover, the mappings*

$$g \rightarrow \tilde{\Psi}_L[g] \text{ and } g \rightarrow \bar{\Psi}_L[g]$$

are both continuous from $\mathcal{M}_2(L)$ to $L_1(\text{Conf}(\mathbb{R}^), \mu_L)$.*

Proof of Proposition 4.15. Let $E_n \subset \mathbb{R}^*$ be a sequence of compact subsets exhausting \mathbb{R}^* and set

$$g_n = 1 + (g - 1)\chi_{E_n}.$$

Clearly, we have $g_n^2 = 1 + (g^2 - 1)\chi_{E_n}$ and

$$g_n^2 \xrightarrow[n \rightarrow \infty]{d.\mathcal{M}_2(L)} g^2. \quad (4.106)$$

Claim: $K(g_n)$ converges to $K(g)$ in the space of locally $\mathcal{L}_{1|2}$ -operators. Indeed, by the block forms of $K(g_n)$ and $K(g)$ as in (4.74), we need to show that for any compact subsets Δ_1, Δ_2 of \mathbb{R}^* such that $\Delta_1 \subset \mathbb{R}_+$ and $\Delta_2 \subset \mathbb{R}_-$, we have

$$\chi_{\Delta_1} g_n V V^* g_n (1 + g_n V V^* g_n)^{-1} \chi_{\Delta_1} \xrightarrow[n \rightarrow \infty]{\text{in trace class}} \chi_{\Delta_1} g V V^* g (1 + g V V^* g)^{-1} \chi_{\Delta_1}; \quad (4.107)$$

$$\chi_{\Delta_2} V^* g_n^2 V (1 + V^* g_n^2 V)^{-1} \chi_{\Delta_2} \xrightarrow[n \rightarrow \infty]{\text{in trace class}} \chi_{\Delta_2} V^* g^2 V (1 + V^* g^2 V)^{-1} \chi_{\Delta_2}; \quad (4.108)$$

$$\chi_{\Delta_1} g_n V (1 + V^* g_n^2 V)^{-1} \chi_{\Delta_2} \xrightarrow[\text{in trace class}]{n \rightarrow \infty} \chi_{\Delta_1} g V (1 + V^* g^2 V)^{-1} \chi_{\Delta_2}. \quad (4.109)$$

Let us prove the first relation (4.107), the proof of second and third relations are similar to that of the first one. First of all,

$$g_n V \xrightarrow[\text{s.o.t.}]{n \rightarrow \infty} g V \text{ and } V^* g_n \xrightarrow[\text{s.o.t.}]{n \rightarrow \infty} V^* g,$$

where s.o.t. stands for the strong operator topology. Hence we have

$$V^* g_n^2 V \xrightarrow[\text{s.o.t.}]{n \rightarrow \infty} V^* g^2 V$$

by continuity of the inverse mapping with respect to strong operator topology (cf. e.g. [Kad68, Lem. 3.2.]), we have

$$(1 + V^* g_n^2 V)^{-1} \xrightarrow[\text{s.o.t.}]{n \rightarrow \infty} (1 + V^* g^2 V)^{-1}.$$

Note also that we have

$$\chi_{\Delta_1} g_n V \xrightarrow[\text{Hilbert-Schmidt}]{n \rightarrow \infty} \chi_{\Delta_1} g V.$$

Combining the above facts and [Grü73, Thm. 1], we obtain that

$$\chi_{\Delta_1} g_n V (1 + V^* g_n^2 V)^{-1} \xrightarrow[\text{Hilbert-Schmidt}]{n \rightarrow \infty} \chi_{\Delta_1} g V (1 + V^* g^2 V)^{-1}.$$

Now by using the following identity

$$\chi_{\Delta_1} g_n V V^* g_n (1 + g_n V V^* g_n)^{-1} \chi_{\Delta_1} = \chi_{\Delta_1} g_n V (1 + V^* g_n^2 V)^{-1} V^* g_n \chi_{\Delta_1}$$

and the triangular inequalities, we conclude the proof of the desired relation (4.107).

As a consequence of our claim, we have the weak convergence of the sequence of measures $\mu_{g_n L g_n}$ to the measure $\mu_{g L g}$. By Proposition 4.10, we also have

$$\mu_{g_n L g_n} = \overline{\Psi}_L[g_n^2] \cdot \mu_L.$$

By Proposition 4.17 and (4.106), $\overline{\Psi}_L[g_n^2]$ converges to $\overline{\Psi}_L[g^2]$ in $L_1(\text{Conf}(\mathbb{R}^*), \mu_L)$. As a consequence, we get the desired relation (4.98). \square

The rest of this section is devoted to the proof of Proposition 4.17.

Lemma 4.18. *Let $f : \mathbb{R}^* \rightarrow \mathbb{C}$ be a Borel function such that $\int_{\mathbb{R}} |f(x)|^4 K_L(x, x) dx < \infty$. Then $f K_L f$ is a Hilbert-Schmidt operator and*

$$\|f K_L f\|_2^2 \leq \int_{\mathbb{R}} |f(x)|^4 K_L(x, x) dx.$$

Proof. We have

$$\begin{aligned}
\|f K_L f\|_2^2 &= \iint_{\mathbb{R}^2} |f(x)|^2 |f(y)|^2 |K_L(x, y)|^2 dx dy \\
&\leq \left(\iint_{\mathbb{R}^2} |f(x)|^4 |K_L(x, y)|^2 dx dy \right)^{1/2} \left(\iint_{\mathbb{R}^2} |f(y)|^4 |K_L(x, y)|^2 dx dy \right)^{1/2} \\
&= \left(\int_{\mathbb{R}} |f(x)|^4 K_L(x, x) dx \right)^{1/2} \left(\int_{\mathbb{R}} |f(y)|^4 K_L(y, y) dy \right)^{1/2} \\
&= \int_{\mathbb{R}} |f(x)|^4 K_L(x, x) dx.
\end{aligned}$$

The proof is complete. \square

Remark 4.14. By definition, if g_1 and g_2 are two functions such that $\log g_1, \log g_2 \in \mathcal{N}(L)$, then

$$\tilde{\Psi}_L[g_1 g_2] = \tilde{\Psi}_L[g_1] \tilde{\Psi}_L[g_2].$$

Lemma 4.19. For any $\varepsilon > 0, M > 0$ so that $\varepsilon < 1 < M$, there exists a constant $C_{\varepsilon, M} > 0$ such that if $g \in \mathcal{M}_2(L)$ satisfies $\varepsilon \leq \inf_{\mathbb{R}} g \leq \sup_{\mathbb{R}} g \leq M$, then

$$\log \mathbb{E}_{\mu_L}(|\tilde{\Psi}_L[g]|^2) \leq C_{\varepsilon, M} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx.$$

Proof. By multiplicativity, it suffices to prove

$$\log \mathbb{E}_{\mu_L}(\tilde{\Psi}_L[g]) \leq C_{\varepsilon, M} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx. \quad (4.110)$$

Since $g^\vee \in \mathcal{M}_2(L)$, by Lemma 4.16, $\log(g^\vee) \in \mathcal{N}(L)$, hence by Remark 4.12, passing to a sequence δ_n if necessary, the functional $\overline{T}_{\log(g^\vee)}$ can be approximated pointwisely by

$$\overline{T}_{(\log(g^\vee))\chi_{\mathbb{R} \setminus (-\delta, \delta)}} = \overline{T}_{\log(g^\vee \chi_{\mathbb{R} \setminus (-\delta, \delta)} + \chi_{[-\delta, \delta]})}.$$

Thus by Fatou's lemma, it suffices to establish (4.110) in the case when $\text{supp}(g - 1)$ is contained in some $\mathbb{R} \setminus (-\delta, \delta)$. In this case, the usual multiplicative functional $\Psi[g]$ is well-defined and we have

$$\tilde{\Psi}_L[g] = \exp(S[\log g] - \mathbb{E}_{\mu_L} S[\log g]) = \frac{\Psi[g]}{\exp(\mathbb{E}_{\mu_L} S[\log g])}.$$

Now by the very definition of determinantal point process $\mu_L = \mathbb{P}_{K_L}$, we have

$$\mathbb{E}_{\mu_L} S_{\log g} = \int_{\mathbb{R}} \log g(x) K(x, x) dx$$

and

$$\mathbb{E}_{\mu_L} \Psi[g] = \det(1 + \sqrt{g-1} K_L \sqrt{g-1}).$$

By [Sim77, Thm. 6.4], if we denote $A = \sqrt{g-1} K_L \sqrt{g-1}$, we have

$$|\det(1 + A) \exp(-\text{tr}(A))| \leq \exp\left(\frac{1}{2} \|A\|_2^2\right).$$

Hence by Lemma 4.18, we have

$$\begin{aligned} \log \mathbb{E}_{\mu_L} \Psi[g] &\leq \text{tr}(A) + \frac{1}{2} \|A\|_2^2 \\ &\leq \int_{\mathbb{R}} (g(x) - 1) K_L(x, x) dx + \frac{1}{2} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx. \end{aligned}$$

An application of (4.104) to the function g yields the existence of a constant $C_{\varepsilon, M} > 0$, such that

$$\left| \int_{\mathbb{R}} \log g(x) K_L(x, x) dx - \int_{\mathbb{R}} (g(x) - 1) K_L(x, x) dx \right| \leq C_{\varepsilon, M} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx.$$

By setting $C'_{\varepsilon, M} = C_{\varepsilon, M} + \frac{1}{2}$, we obtain

$$\log \mathbb{E}_{\mu_L} \tilde{\Psi}_L[g] = \log \mathbb{E}_{\mu_L} \Psi[g] - \mathbb{E}_{\mu_L} S[\log g] \leq C'_{\varepsilon, M} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx.$$

The proof is complete. \square

Lemma 4.20. *Let $\varepsilon > 0$, $M > 0$ be two positive numbers such that $\varepsilon < 1 < M$. There exists a constant $C > 0$ depending on ε , M , such that if $g_1, g_2 \in \mathcal{M}_2(L)$ satisfy*

$$\varepsilon \leq \inf_{\mathbb{R}} g_1 \leq \sup_{\mathbb{R}} g_1 \leq M, \quad \varepsilon \leq \inf_{\mathbb{R}} g_2 \leq \sup_{\mathbb{R}} g_2 \leq M,$$

then we have

$$\frac{(\mathbb{E}_{\mu_L} |\tilde{\Psi}_L[g_1] - \tilde{\Psi}_L[g_2]|)^2}{\mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g_1]|^2)} \leq \exp \left(C \int_{\mathbb{R}} |g_1(x) - g_2(x)|^2 K_L(x, x) dx \right) - 1.$$

Proof. Set $g = g_2/g_1$. Since $\tilde{\Psi}_L[g_1] - \tilde{\Psi}_L[g_2] = \tilde{\Psi}_L[g_1](1 - \tilde{\Psi}_L[g])$, we have

$$(\mathbb{E}_{\mu_L} |\tilde{\Psi}_L[g_1] - \tilde{\Psi}_L[g_2]|)^2 \leq \mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g_1]|^2) \cdot \mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g] - 1|^2). \quad (4.111)$$

By the inequality (4.95), we have

$$\mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g] - 1|^2) = \mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g]|^2) - 2\mathbb{E}_{\mu_L} |\tilde{\Psi}_L[g]| + 1 \leq \mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g]|^2) - 1. \quad (4.112)$$

Since $\varepsilon/M \leq \inf_{\mathbb{R}} g \leq \sup_{\mathbb{R}} g \leq M/\varepsilon$, by Lemma 4.19, there exists $C_{\varepsilon, M} > 0$, such that

$$\mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g]|^2) \leq \exp \left(C_{\varepsilon, M} \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx \right).$$

Hence there exists $C'_{\varepsilon, M} > 0$, such that

$$\mathbb{E}_{\mu_L} (|\tilde{\Psi}_L[g]|^2) \leq \exp \left(C'_{\varepsilon, M} \int_{\mathbb{R}} |g_1(x) - g_2(x)|^2 K_L(x, x) dx \right). \quad (4.113)$$

Substituting the inequalities (4.112) and (4.113) into (4.111), we obtain the desired inequality. \square

4.8 Proof of Theorem 4.9

Proof of Theorem 4.9. By Theorem 4.14, it suffices to check the inequality (4.96) under the assumption of Theorem 4.14. Indeed, we have

$$\begin{aligned} & \int_{\mathbb{R}} |g(x) - 1|^2 K_L(x, x) dx \\ &= \int_{|x| \geq \varepsilon} |g(x) - 1|^2 K_L(x, x) dx + \int_{|x| < \varepsilon} |g(x) - 1|^2 K_L(x, x) dx \\ &=: I + II. \end{aligned}$$

The relation $I < \infty$ follows from the boundedness of g and the assumption (4.75). For the second term, we have

$$II \leq C^2 \int_{|x| < \varepsilon} x^2 K_L(x, x) dx < \infty.$$

This proof of Theorem 4.9 is complete. \square

4.9 Proof of Theorem B

By [Ols, Thm. 2.4] and [BO00, §6], if we assume that

$$\left| \frac{z + z'}{2} \right| < \frac{1}{2},$$

then the Whittaker kernel $\mathcal{K} = \mathcal{K}_{z, z'}$ admits a bounded L -operator as in (1.8), such that the bounded operator $V : L^2(\mathbb{R}_-) \rightarrow L^2(\mathbb{R}_+)$ has as kernel:

$$\frac{\sin \pi z \sin \pi z'}{\pi^2} \frac{\left(\frac{x}{-y} \right)^{\frac{z+z'}{2}} e^{-\frac{x-y}{2}}}{x-y}, \text{ where } x > 0, y < 0.$$

In other words, the L -kernel $\mathcal{L}(x, y) = \mathcal{L}_{z, z'}(x, y)$ of the kernel $\mathcal{K}(x, y) = \mathcal{K}_{z, z'}(x, y)$ is given by

$$\mathcal{L}_{z, z'}(x, y) = \frac{\mathcal{A}^+(x) \mathcal{A}^-(y) + \mathcal{A}^-(x) \mathcal{A}^+(y)}{x - y}, \quad x, y \in \mathbb{R}^*, \quad (4.114)$$

where

$$\mathcal{A}(x) = \frac{\sqrt{\sin \pi z \sin \pi z'}}{\pi} |x|^{\operatorname{sgn}(x) \frac{z+z'}{2}} e^{-\frac{|x|}{2}}, \text{ where } x \neq 0.$$

This function \mathcal{A} satisfies the following conditions:

- the support of \mathcal{A} in \mathbb{R}^* is the whole punctured line \mathbb{R}^* ;
- $\mathcal{A} \in C^\infty(\mathbb{R}^*) \cap L^2(\mathbb{R})$.

Thus we have shown the following

Lemma 4.21. *If $|z + z'| < 1$, then the L -kernel $\mathcal{L} = \mathcal{L}_{z,z'}$ in (4.114) satisfies Condition 3.*

Recall that if $\mathbf{p} = (p_1^+, \dots, p_n^+; p_1^-, \dots, p_n^-)$, then we set

$$g_{\mathbf{p}}(x) = \prod_{i=1}^n \left(\frac{x - p_i^+}{x - p_i^-} \chi_{\{x>0\}} + \frac{x - p_i^-}{x - p_i^+} \chi_{\{x<0\}} \right).$$

Let $\lambda := \frac{|p_1^- \cdots p_n^-|}{p_1^+ \cdots p_n^+}$ and recall the formula (4.83): $c_\lambda = \lambda \chi_{\mathbb{R}_+} + \lambda^{-1} \chi_{\mathbb{R}_-}$. Set

$$h_{\mathbf{p}}(x) = c_\lambda(x) |g_{\mathbf{p}}(x)| = c_\lambda(x) \operatorname{sgn}(g_{\mathbf{p}}(x)) g_{\mathbf{p}}(x).$$

That is,

$$h_{\mathbf{p}}(x) = \prod_{i=1}^n \left| \frac{x/p_i^+ - 1}{x/p_i^- - 1} \chi_{\{x>0\}} + \frac{x/p_i^- - 1}{x/p_i^+ - 1} \chi_{\{x<0\}} \right|.$$

The proof of the following lemma is immediate.

Lemma 4.22. *The function $h_{\mathbf{p}}$ is bounded and there exists $C > 0$ such that*

$$|h_{\mathbf{p}}(x) - 1| \leq C|x|.$$

Moreover, for any $\varepsilon > 0$, the subset $\{x \in \mathbb{R}^* : |h_{\mathbf{p}}(x) - 1| > \varepsilon\}$ is away from 0.

Proof of Proposition 1.2 and Theorem B. By Proposition 4.1, we have $\mathcal{P}_{z,z'}^{\mathbf{p}} = \mu_{g_{\mathbf{p}} \mathcal{L} g_{\mathbf{p}}}$. By Lemma 4.11, we have $\mu_{g_{\mathbf{p}} \mathcal{L} g_{\mathbf{p}}} = \mu_{|g_{\mathbf{p}}| \mathcal{L} |g_{\mathbf{p}}|} = \mu_{h_{\mathbf{p}} \mathcal{L} h_{\mathbf{p}}}$. Finally, by Theorem 4.9 and Lemma 4.22, the following limit

$$\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2](X) := \lim_{\delta \rightarrow 0^+} \left(\sum_{x \in X, |x| \geq \delta} \log h_{\mathbf{p}}(x)^2 - \mathbb{E}_{\mu_{\mathcal{L}}} \sum_{x \in X, |x| \geq \delta} \log h_{\mathbf{p}}(x)^2 \right)$$

exists for $\mu_{\mathcal{L}}$ -almost every configuration $X \in \operatorname{Conf}(\mathbb{R}^*)$. Moreover, the function

$$X \mapsto \exp(\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2](X))$$

is in $L^1(\operatorname{Conf}(\mathbb{R}^*), \mu_{\mathcal{L}})$ and we have

$$\mu_{h_{\mathbf{p}} \mathcal{L} h_{\mathbf{p}}}(dX) = \frac{\exp(\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2](X))}{\mathbb{E}_{\mu_{\mathcal{L}}} \left[\exp(\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2]) \right]} \cdot \mu_{\mathcal{L}}(dX),$$

that is,

$$\mathcal{P}_{z,z'}^{\mathbf{p}}(dX) = \frac{\exp(\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2](X))}{\mathbb{E}_{\mu_{\mathcal{L}}} \left[\exp(\overline{\mathcal{S}}_{\mathcal{L}}[\log h_{\mathbf{p}}^2]) \right]} \cdot \mathcal{P}_{z,z'}(dX).$$

□

5 Appendix

Proof of Proposition 2.2. By homogeneity, we may assume, without loss of generality, that $\|A\|_{\mathcal{L}_{1|2}} \leq 1$ and $\|B\|_{\mathcal{L}_{1|2}} \leq 1$. Write A and B in block forms:

$$A = \begin{bmatrix} a_1 & b_1 \\ c_1 & d_1 \end{bmatrix}, \quad B = \begin{bmatrix} a_2 & b_2 \\ c_2 & d_2 \end{bmatrix},$$

then we have

$$AB = \begin{bmatrix} a_1a_2 + b_1c_2 & a_1b_2 + b_1d_2 \\ c_1a_2 + d_1c_2 & c_1b_2 + d_1d_2 \end{bmatrix}.$$

By applying the operator ideal property $\|ab\|_1 \leq \|a\|_1\|b\|$, $\|ab\|_1 \leq \|a\|_2\|b\|$ and the Hölder inequality $\|ab\|_1 \leq \|a\|_2\|b\|_2$, we get

$$\begin{aligned} \|AB\|_{\mathcal{L}_{1|2}} &= \|a_1a_2 + b_1c_2\|_1 + \|c_1b_2 + d_1d_2\|_1 + \|a_1b_2 + b_1d_2\|_2 + \|c_1a_2 + d_1c_2\|_2 \\ &\leq \|a_1\|_1\|a_2\| + \|b_1\|_2\|c_2\|_2 + \|c_1\|_2\|b_2\|_2 + \|d_1\|_1\|d_2\| \\ &\quad + \|a_1\|_1\|b_2\| + \|b_1\|_2\|d_2\| + \|c_1\|_2\|a_2\| + \|d_1\|_1\|c_2\|_2 \\ &\leq \|a_1\|_1 + \|b_1\|_2 + \|c_1\|_2 + \|d_1\|_1 + \|a_1\|_1 + \|b_1\|_2 + \|c_1\|_2 + \|d_1\|_1 \\ &\leq 2(\|a_1\|_1 + \|b_1\|_2 + \|c_1\|_2 + \|d_1\|_1) \leq 2. \end{aligned}$$

□

Proof of Proposition 2.3. The proof is easy from the definition of $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$ and the ideal property of trace-class and Hilbert-Schmidt class. □

Proof of Proposition 2.4. By the relation (2.21), under the hypothesis of Proposition 2.4 on A, B , the two operators A, B are both in $\mathcal{L}_2(L^2(\mathbb{R}))$, hence $AB \in \mathcal{L}_1(L^2(\mathbb{R}))$. By the ideal property of $\mathcal{L}_1(L^2(\mathbb{R}))$, the operator $(1 + A)^{-1}AB$ belongs to $\mathcal{L}_1(L^2(\mathbb{R}))$ and hence belongs to $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. We can write

$$(1 + A)^{-1}B = (1 + A)^{-1}((1 + A)B - AB) = B - (1 + A)^{-1}AB,$$

hence the operator $(1 + A)^{-1}B$ belongs to $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. Similar argument yields the fact that the operator $B(1 + A)^{-1}$ also belongs to $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. □

Proof of Proposition 2.5. Fix a pair of operators A, B in $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$. Note first that by Proposition 2.2, the operator $A + B + AB$ is in the space $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$, hence the extended Fredholm determinant $\det((1 + A)(1 + B)) = \det(1 + A + B + AB)$ is well-defined. By the multiplicativity property of the usual Fredholm determinant, the desired identity holds whenever $A, B \in \mathcal{L}_1(L^2(\mathbb{R}))$, see, e.g. [Sim77, Thm. 3.8]. Thus by the continuity of the function $A \mapsto \det(1 + A)$ on $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$, for proving the desired identity, it suffices

to show that there exist two sequences $(A_n)_{n \in \mathbb{N}}$ and $(B_n)_{n \in \mathbb{N}}$ in $\mathcal{L}_1(L^2(\mathbb{R}))$ such that we have the following convergences in the space $\mathcal{L}_{1|2}(L^2(\mathbb{R}))$:

$$A_n \xrightarrow{n \rightarrow \infty} A, \quad B_n \xrightarrow{n \rightarrow \infty} B \quad \text{and} \quad A_n B_n \xrightarrow{n \rightarrow \infty} AB. \quad (5.115)$$

To this end, take any two sequences $(P_n)_{n \in \mathbb{N}}$ and $(Q_n)_{n \in \mathbb{N}}$ of finite rank orthogonal projections on $L^2(\mathbb{R}_+)$ and $L^2(\mathbb{R}_-)$ respectively, assume that P_n and Q_n converge in the strong operator topology to the orthogonal projections P_+ and P_- respectively. Now we may set

$$A_n = (P_n + Q_n)A, \quad B = B(P_n + Q_n).$$

Then it is clear that the finite rank operators A_n and B_n satisfy all the desired conditions in (5.115). Note that we intentionally obtain A_n and B_n by multiplying $P_n + Q_n$ on the left side of A and on the right side of B , so that the third condition in (5.115) is satisfied. \square

Proof of Proposition 2.6. From Grothendieck's definition of Fredholm determinant:

$$\det(1 + T) = \sum_{k=0}^{\infty} \text{tr}(\wedge^k(T)), \quad T \in \mathcal{L}_1(L^2(\mathbb{R})),$$

and the fact that, once $A \in \mathcal{L}_1(L^2(\mathbb{R}))$ and f is a bounded function, then

$$\text{tr}(\wedge^k(fA)) = \text{tr}(\wedge^k(M_f) \circ \wedge^k(A)) = \text{tr}(\wedge^k(A) \circ \wedge^k(M_f)) = \text{tr}(\wedge^k(Af)),$$

we see that the identity (2.24) holds when $A \in \mathcal{L}_1(L^2(\mathbb{R}))$. For $A \in \mathcal{L}_{1|2}(L^2(\mathbb{R}))$, we may argue similarly as in the proof of Proposition 2.5. See also [Buf13] for the proof in more general case. \square

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